

Anderson-Darling Type Goodness-of-fit Statistic Based on a Multifold Integrated Empirical Distribution Function

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- Summary

I. Anderson-Darling statistic and its extension

Goodness-of-fit tests

- ▶ X_1, \dots, X_n : i.i.d. sequence from cdf F
- ▶ Goodness-of-fit test:

$$H_0 : F = G \quad \text{vs.} \quad H_1 : F \neq G$$

(G is a given cdf)

- ▶ When G is continuous, we can assume $G(x) = x$ (i.e., $\text{Unif}(0,1)$) WLOG.
- ▶ Empirical distribution function

$$F_n(x) = \frac{1}{n} \sum_{i=1}^n \mathbf{1}(X_i \leq x)$$

- ▶ Test statistic is defined as a measure of discrepancy between $F_n(x)$ and $G(x) = x$.

Goodness-of-fit tests (cont)

- ▶ We focus on two integral-type test statistics.
- ▶ Anderson-Darling (1952) statistic:

$$A_n = n \int_0^1 \frac{1}{x(1-x)} (F_n(x) - x)^2 dx$$

- ▶ Watson (1961) statistic
(for testing uniformity on the unit sphere in \mathbb{R}^2):

$$U_n = n \int_0^1 (F_n(x) - x)^2 dx - n \left\{ \int_0^1 (F_n(x) - x) dx \right\}^2$$

- ▶ Limiting null distributions:

Let ξ_1, ξ_2, \dots be i.i.d. sequence from $N(0, 1)$. As $n \rightarrow \infty$,

$$A_n \xrightarrow{d} \sum_{k=1}^{\infty} \frac{1}{k(k+1)} \xi_k^2, \quad U_n \xrightarrow{d} \sum_{k=1}^{\infty} \frac{1}{2\pi^2 k^2} (\xi_{2k-1}^2 + \xi_{2k}^2)$$

Closely looking at Anderson-Darling

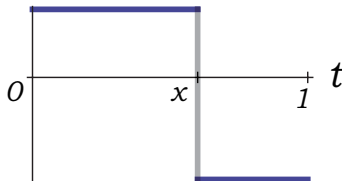
- ▶ Anderson-Darling statistic

$$A_n = \int_0^1 \frac{B_n(x)^2}{x(1-x)} dx \quad \text{where } B_n(x) = \sqrt{n}(F_n(x) - x)$$

Here,

$$B_n(x) = \sqrt{n} \int_0^1 h^{[0]}(t; x) dF_n(t), \quad h^{[0]}(t; x) = \mathbf{1}(t \leq x) - x$$

- ▶ $h^{[0]}(t; x)$



An extension to Anderson-Darling

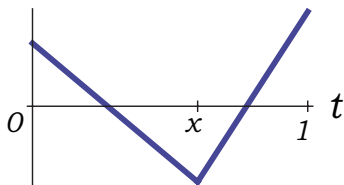
- ▶ To propose a new class of test statistics, instead of $h^{[0]}(\cdot; x)$, we prepare different type $h^{[m]}(\cdot; x)$.
- ▶ Note first that $h^{[0]}(\cdot; x)$ is piecewise constant s.t.

$$\int_0^1 h^{[0]}(t; x) \cdot 1 dt = 0$$

- ▶ Define $h^{[1]}(\cdot; x)$ to be continuous and piecewise linear s.t.

$$\int_0^1 h^{[1]}(t; x) \cdot (at + b) dt = 0, \quad \forall a, b$$

- ▶ $h^{[1]}(\cdot; x)$



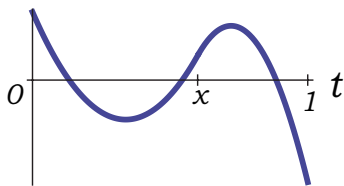
An extension to Anderson-Darling (cont)

- ▶ General form of $h^{[m]}(t; x)$:

$$h^{[m]}(t; x) = \frac{1}{m!} (x - t)^m \mathbb{1}(t \leq x) - \sum_{k=0}^m \int_0^x \frac{1}{m!} (x - u)^m L_k(u) du \times L_k(t)$$

where $L_k(\cdot)$ is the Legendre polynomial of degree k

- ▶ $h^{[2]}(\cdot; x)$



An extension to Anderson-Darling (cont)

- ▶ We propose an extension of Anderson-Darling:

$$A_n^{[m]} = \int_0^1 \frac{B_n^{[m]}(x)^2}{\{x(1-x)\}^{m+1}} dx$$

where

$$\begin{aligned} B_n^{[m]}(x) &= \sqrt{n} \int_0^1 h^{[m]}(t; x) dF_n(t) \\ &= \sqrt{n} \left\{ \int_{x > x_m > \dots > x_1 > 0} \dots \int F_n(x_1) dx_1 \dots dx_m \right. \\ &\quad \left. - \sum_{k=0}^m \int_{x > x_m > \dots > x_1 > 0} \dots \int L_k(x_1) dx_1 \dots dx_m \int_0^1 L_k(t) dF_n(t) \right\} \end{aligned}$$

(m -fold integral of empirical distribution function)

- ▶ $A_n^{[m]}$ is well-defined (the integral exists) whenever $X_i \in (0, 1)$.
- ▶ $A_n^{[0]}$ is the original Anderson-Darling.

II. Limiting null distribution

Main results — Limiting null distribution

- ▶ $W(\cdot)$: the Winer process on $[0, 1]$
 $B(x) = W(x) - xW(1)$: Brownian bridge
- ▶ Let $B_n^{[m]}(x) = \int_0^1 h^{[m]}(t; x) dB_n(x)$.
We can prove that as $n \rightarrow \infty$,

$$B_n^{[m]}(\cdot) \xrightarrow{d} B^{[m]}(\cdot) \text{ in } L^2, \text{ where } B^{[m]}(x) = \int_0^1 h^{[m]}(t; x) dB(t)$$

and hence (by continuous mapping)

$$A_n^{[m]} = \int_0^1 \frac{B_n^{[m]}(x)^2}{\{x(1-x)\}^{m+1}} dx \xrightarrow{d} A^{[m]} := \int_0^1 \frac{B^{[m]}(x)^2}{\{x(1-x)\}^{m+1}} dx$$

- ▶ We will examine these limiting distributions $B^{[m]}(\cdot)$ and $A^{[m]}$.

Main results — Limiting null distribution (cont)

Theorem (Karhunen-Loève expansion)

$$\frac{B^{[m]}(x)}{\{x(1-x)\}^{(m+1)/2}} = \sum_{k=m+1}^{\infty} \sqrt{\frac{(k-m-1)!}{(k+m+1)!}} L_k^{(m+1)}(x) \xi_k$$

(uniformly in x , with prob. 1), where

$$\xi_k = \int_0^1 L_k(t) dB(t), \quad i.i.d. N(0, 1)$$

$L_k^{(m+1)}$ is the associate Legendre function. □

Corollary (Limiting null distribution of $A_n^{[m]}$)

$$A^{[m]} = \int_0^1 \frac{B^{[m]}(x)^2}{\{x(1-x)\}^{m+1}} dx = \sum_{k=m+1}^{\infty} \frac{(k-m-1)!}{(k+m+1)!} \xi_k^2$$

$\xi_k^2 \sim \chi^2(1)$ i.i.d.

III. Moment generating function

Moment generating function

- ▶ The moment generating function (Laplace transform) of

$$A^{[m]} = \sum_{k=1}^{\infty} \frac{1}{\lambda_k} \xi_k^2, \quad \lambda_k = k(k+1) \cdots (k+2m+1)$$

is

$$E\left[e^{sA^{[m]}}\right] = \prod_{k=1}^{\infty} \left(1 - \frac{2s}{\lambda_k}\right)^{-\frac{1}{2}}$$

Theorem

Let $x_j(s)$ ($j = 0, 1, \dots, 2m+1$) be the solution of

$$\lambda_x - 2s = 0, \text{ i.e., } x(x+1) \cdots (x+2m+1) - 2s = 0$$

Then

$$E\left[e^{sA^{[m]}}\right] = \prod_{j=0}^{2m+1} \sqrt{\frac{\Gamma(1 - x_j(s))}{j!}}$$



Moment generating function ($m = 0$)

- ▶ When $m = 0$ (Anderson-Darling), $\lambda_k = k(k + 1)$

The equation

$$x(x + 1) - 2s = 0$$

has a solution

$$x_0(s), x_1(s) = -\frac{1}{2} \pm \sqrt{1 + 8s}$$

- ▶ Hence,

$$\begin{aligned} E \left[e^{sA^{[0]}} \right] &= \sqrt{\Gamma(1 - x_0(s))\Gamma(1 - x_1(s))} \\ &= \sqrt{\frac{2\pi s}{-\cos \frac{\pi}{2} \sqrt{1 + 8s}}} \end{aligned}$$

(Anderson and Darling, 1952)

- ▶ Euler's reflection formula $\Gamma(z)\Gamma(1 - z) = \pi / \sin(\pi z)$ is used.

Moment generating function ($m = 1$)

- ▶ When $m = 1$. $\lambda_k = k(k+1)(k+2)(k+3)$.
- ▶ The equation

$$x(x+1)(x+2)(x+3) - 2s = 0$$

has the explicit solution, because by letting $x = y - 3/2$,

$$\begin{aligned} \text{LHS} &= (y - 3/2)(y - 1/2)(y + 1/2)(y + 3/2) - 2s \\ &= \{y^2 - (3/2)^2\} \{y^2 - (1/2)^2\} - 2s \end{aligned}$$

is a quadratic equation in $y^2 = (x + 3/2)^2$.

- ▶ As a result,

$$x_0(s), x_1(s), x_2(s), x_3(s) = \frac{1}{2} \left(\pm \sqrt{5 \pm 4\sqrt{2s+1}} - 3 \right),$$

$$E \left[e^{sA^{[1]}} \right] = \frac{\pi s}{\sqrt{3 \cos\left(\frac{\pi}{2} \sqrt{5 - 4\sqrt{1+2s}}\right) \cos\left(\frac{\pi}{2} \sqrt{5 + 4\sqrt{1+2s}}\right)}}$$

Moment generating function ($m = 2$)

- ▶ When $m = 2$,

$$E \left[e^{sA^{[2]}} \right] = \frac{(\pi s)^{3/2}}{\sqrt{-4320 \cos(\pi \sqrt{\eta_1}) \cosh(\pi \sqrt{\eta_2}) \cosh(\pi \sqrt{\eta_3})}}$$

where

$$\eta = \sqrt[3]{27s + 80 + 3\sqrt{81s^2 + 480s - 1728}}$$

and

$$\eta_1 = \frac{1}{12\eta} (4\eta^2 + 35\eta + 112)$$

$$\eta_2 = \frac{1}{12\eta} (4e^{-\pi i/3} \eta^2 - 35\eta + 112e^{\pi i/3})$$

$$\eta_3 = \frac{1}{12\eta} (4e^{\pi i/3} \eta^2 - 35\eta + 112e^{-\pi i/3})$$

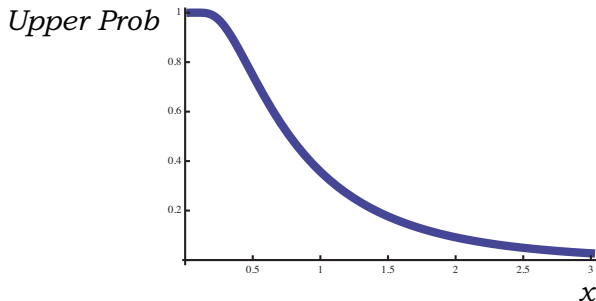
Calculation of upper prob.

- ▶ Finite representation is useful in numerical calculation.

$$P(A^{[m]} > x) = \sum_{k=1}^{\infty} \frac{(-1)^{k-1}}{\pi} \int_{\lambda_{2k-1}/2}^{\lambda_{2k}/2} \frac{e^{-xs}}{s \sqrt{|E[e^{sA^{[m]}]}|}} ds$$

(Smirnov-Slepian technique, see Slepian (1958)).

- ▶ The case $m = 0$:



IV. Statistical power

Statistical power

- ▶ We have the sample counterpart of the KL-expansion:

$$\frac{B_n^{[m]}(x)}{\{x(1-x)\}^{(m+1)/2}} = \sum_{k \geq m+1} \sqrt{\frac{(k-m-1)!}{(k+m+1)!}} L_k^{(m+1)}(x) \hat{\xi}_k$$

where

$$\hat{\xi}_k = \int_0^1 L_k(x) dB_n(x) = \frac{1}{\sqrt{n}} \sum_{i=1}^n L_k(X_i), \quad k \geq 1$$

- ▶ The (extended) Anderson-Darling statistics are also written in terms of $\hat{\xi}_k$'s as

$$A_n^{[0]} = \sum_{k \geq 1} \frac{1}{k(k+1)} \hat{\xi}_k^2 = \frac{1}{2} \hat{\xi}_1^2 + \frac{1}{6} \hat{\xi}_2^2 + \frac{1}{12} \hat{\xi}_3^2 + \dots$$

$$A_n^{[1]} = \sum_{k \geq 2} \frac{1}{(k-1)k(k+1)(k+2)} \hat{\xi}_k^2 = \frac{1}{24} \hat{\xi}_2^2 + \frac{1}{120} \hat{\xi}_3^2 + \dots$$

Statistical power (cont)

- ▶ First two components:

$$\hat{\xi}_1 = \sqrt{12nm_1}, \quad \hat{\xi}_2 = 6\sqrt{5n} \times (m_2 - 1/12),$$

where $m_k = \frac{1}{n} \sum_{i=1}^n (X_i - 1/2)^k$ (the sample k th moment around $1/2$)

- ▶ $A_n^{[0]} = \hat{\xi}_1^2/2 + \dots$ has much power for mean-shift alternative, and
- ▶ $A_n^{[1]} = \hat{\xi}_2^2/24 + \dots$ has much power for dispersion-change alternative

V. Extension of Watson's statistic

Watson's statistic and its extension

- ▶ Similar extensions are possible for Watson's statistic. Let

$$U_n^{[m]} = \int_0^1 C_n^{[m]}(x)^2 dx, \quad C_n^{[m]} = \int_0^1 h^{[m]}(t; x) dF_n(t),$$

where

$$h^{[m]}(t; x) = \frac{(t-x)^m}{m!} \mathbf{1}(t-x \leq 0) + \frac{1}{(m+1)!} b_{m+1}(t-x)$$

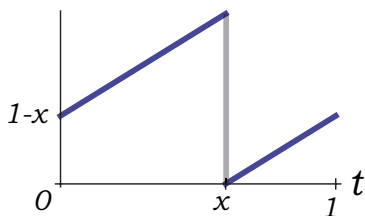
- ▶ $b_m(y)$ is the **Bernoulli polynomial**, which satisfies

$$b_m(y+1) = b_m(y) + my^{m-1}.$$

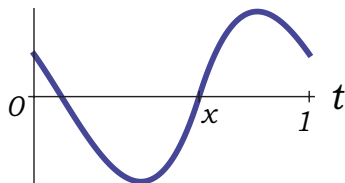
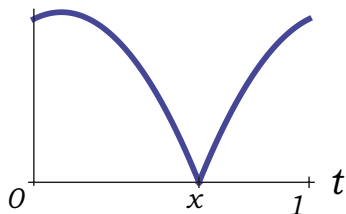
- ▶ $U_n^{[0]}$ is the original Watson statistic.
- ▶ $U_n^{[1]}$ is proposed by Henze and Nikitin (2002).

Watson's statistic and its extension (cont)

▶ $h^{[0]}(\cdot; x)$



▶ $h^{[1]}(\cdot; x)$ and $h^{[2]}(\cdot; x)$



Limiting null distribution

- ▶ Let

$$C_n^{[m]}(\cdot) \xrightarrow{d} C^{[m]}(\cdot) \quad \text{and} \quad U_n^{[m]} \xrightarrow{d} U^{[m]}.$$

- ▶ KL-expansion of $C^{[m]}(x)$:

$$C^{[m]}(x) = \sum_{k=1}^{\infty} \frac{1}{(2k\pi)^{m+1}} \left\{ l_{2k-1}^{[m]}(x) \xi_{2k-1} + l_{2k}^{[m]}(x) \xi_{2k} \right\},$$

where

$$l_{2k-1}^{[m]}(x) = \sin\left(2k\pi x - \frac{m+1}{2}\pi\right), \quad l_{2k}^{[m]}(x) = \cos\left(2k\pi x - \frac{m+1}{2}\pi\right),$$

$$\xi_{2k-1} = \int_0^1 \sin(2k\pi x) dB(x), \quad \xi_{2k} = \int_0^1 \cos(2k\pi x) dB(x).$$

- ▶ Consequently,

$$U^{[m]} = \sum_{k=1}^{\infty} \frac{1}{(2k\pi)^{2(m+1)}} \{ \xi_{2k-1}^2 + \xi_{2k}^2 \}.$$

Summary

- ▶ We proposed a class of extended Anderson-Darling statistics $A_n^{[m]}$ ($m \geq 0$) based on m -fold integrated empirical distribution function.
- ▶ The limiting null distribution $A^{[m]}$ is explicitly derived as weighted infinite sums of chi-square random variables.
- ▶ We provided moment generating function of $A^{[m]}$ without using infinite product.
- ▶ The same-type extension for Watson's statistic U_n is possible.
- ▶ Acknowledgment: The authors thank Y. Nishiyama of ISM.

References

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