

Ergodic Convergence Rates of Markov Processes

—Eigenvalues, Inequalities and Ergodic Theory

Mu-Fa Chen

(Beijing Normal University)
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- Introduction.
- Variational formulas of the first eigenvalue by couplings.
- Estimates of inequalities' constant by generalized Cheeger's method.
- A diagram of nine types of ergodicity and a table of explicit criteria in dimension one.

I. Introduction

1.1 Definition. The first (non-trivial) eigenvalue:

$$Q = \begin{pmatrix} -b_0 & b_0 & 0 & 0 & \cdots \\ a_1 & -(a_1+b_1) & b_1 & 0 & \cdots \\ 0 & a_2 & -(a_2+b_2) & b_2 & \cdots \\ \vdots & \vdots & \ddots & \ddots & \ddots \end{pmatrix}$$

$$a_i > 0, b_i > 0.$$

$Q1 = 0$. Trivial eigenvalue: $\lambda_0 = 0$.

Question: Next eigenvalue of $-Q$: $\lambda_1 = ?$

Elliptic operator in \mathbb{R}^d ; Laplacian on Riemannian manifolds. Importance: leading term.

1.2 Difficulties

Example 1: Trivial case(two points). Two parameters.

$$\begin{pmatrix} -b & b \\ a & -a \end{pmatrix}, \quad \lambda_1 = a + b.$$

λ_1 is increasing in each of the parameters!

Example 2: Three points. Four parameters.

$$\begin{pmatrix} -b_0 & b_0 & 0 \\ a_1 & -(a_1 + b_1) & b_1 \\ 0 & a_2 & -a_2 \end{pmatrix},$$

$$\lambda_1 = 2^{-1} \left[a_1 + a_2 + b_0 + b_1 - \sqrt{(a_1 - a_2 + b_0 - b_1)^2 + 4a_1 b_1} \right].$$

Example 3: Four points.

Six parameters: $b_0, b_1, b_2, a_1, a_2, a_3$.

$$\lambda_1 = \frac{D}{3} - \frac{C}{3 \cdot 2^{1/3}} + \frac{2^{1/3} (3B - D^2)}{3C},$$

where

$$D = a_1 + a_2 + a_3 + b_0 + b_1 + b_2,$$

$$B = a_3 b_0 + a_2 (a_3 + b_0) + a_3 b_1 + b_0 b_1 + b_0 b_2 \\ + b_1 b_2 + a_1 (a_2 + a_3 + b_2),$$

$$C = \left(A + \sqrt{4(3B - D^2)^3 + A^2} \right)^{1/3},$$

$$\begin{aligned}
A = & -2a_1^3 - 2a_2^3 - 2a_3^3 + 3a_3^2b_0 + 3a_3b_0^2 - \\
& 2b_0^3 + 3a_3^2b_1 - 12a_3b_0b_1 + 3b_0^2b_1 + 3a_3b_1^2 + \\
& 3b_0b_1^2 - 2b_1^3 - 6a_3^2b_2 + 6a_3b_0b_2 + 3b_0^2b_2 + \\
& 6a_3b_1b_2 - 12b_0b_1b_2 + 3b_1^2b_2 - 6a_3b_2^2 + 3b_0b_2^2 + \\
& 3b_1b_2^2 - 2b_2^3 + 3a_1^2(a_2 + a_3 - 2b_0 - 2b_1 + b_2) + \\
& 3a_2^2[a_3 + b_0 - 2(b_1 + b_2)] + 3a_2[a_3^2 + b_0^2 - 2b_1^2 - \\
& b_1b_2 - 2b_2^2 - a_3(4b_0 - 2b_1 + b_2) + 2b_0(b_1 + b_2)] + \\
& 3a_1[a_2^2 + a_3^2 - 2b_0^2 - b_0b_1 - 2b_1^2 - a_2(4a_3 - 2b_0 + \\
& b_1 - 2b_2) + 2b_0b_2 + 2b_1b_2 + b_2^2 + 2a_3(b_0 + b_1 + b_2)].
\end{aligned}$$

The role of each parameter is completely mazed!
Not solvable when space has more than five points!
Conclusion: Impossible to compute λ_1 explicitly!

Perturbation of eigenvalues

Example: Infinite triangle matrix
(Birth-death processes).

$b_i (i \geq 0)$	$a_i (i \geq 1)$	λ_1	degree of eigenfun.
$i + \beta$ $(\beta > 0)$	$2i$	1	1
$i + 1$	$2i + 3$		
$i + 1$	$2i + (4 + \sqrt{2})$		

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$i + 1$	$2i + 3$	2	
$i + 1$	$2i + (4 + \sqrt{2})$	3	

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$i + 1$	$2i + 3$	2	2
$i + 1$	$2i + (4 + \sqrt{2})$	3	3

Sensitive. In general, it is too hard to estimate λ_1 !

II. New variational formulas of the first eigenvalue

2.1 Story of the study on λ_1 in geometry

(M, g) : compact Riemannian manifold.

Discrete spectrum: $0 = \lambda_0 < \lambda_1 < \dots$.

- g : Riemannian metric.
- d : dimension. \mathbb{S}^d : $D = \pi$, $\text{Ric} = d - 1$.
- D : diameter.
- $\text{Ricci}_M \geq K g$ for some $K \in \mathbb{R}$.

Idea: Use geometric quantities d , D and K to estimate λ_k 's of Laplacian Δ .

Five books!

- Chavel, I. (1984): Eigenvalues in Riemannian Geometry, Academic Press
- Bérard, P. H. (1986): Spectral Geometry: Direct and Inverse Problem, LNM. vol. 1207, Springer-Verlag. Including 2000 references.
- Schoen, R. and Yau, S. T. (1988): Differential Geometry (In Chinese), Science Press, Beijing, China
- Li, P. (1993): Lecture Notes on Geometric Analysis, Seoul National U., Korea
- Ma, C. Y. (1993): The Spectrum of Riemannian Manifolds (In Chinese), Press of Nanjing U., Nanjing

Eight of the most beautiful lower bounds:

Case 1: $K \geq 0$

Author(s)	Lower bound
A. Lichnerowicz (1958)	$\frac{d}{d-1} K$
P. H. Bérard, G. Besson & S. Gallot (1985)	$d \left\{ \frac{\int_0^{\pi/2} \cos^{d-1} t dt}{\int_0^{D/2} \cos^{d-1} t dt} \right\}^{2/d}$ $K = d - 1$
P. Li & S. T. Yau (1980)	$\frac{\pi^2}{2 D^2}$
J. Q. Zhong & H. C. Yang (1984)	$\frac{\pi^2}{D^2}$

Case 2: $K \leq 0$

Author(s)	Lower bound
P. Li & S. T. Yau (1980)	1 <hr/> $D^2(d-1) \exp [1 + \sqrt{1 + 16\alpha^2}]$
K. R. Cai (1991)	$\frac{\pi^2}{D^2} + K$
H. C. Yang (1989) & F. Jia (1991)	$\frac{\pi^2}{D^2} e^{-\alpha}, \quad \text{if } d \geq 5$
H. C. Yang (1989) & F. Jia (1991)	$\frac{\pi^2}{2D^2} e^{-\alpha'}, \quad \text{if } 2 \leq d \leq 4$

$$\begin{aligned}\alpha &= D \sqrt{|K|(d-1)/2}, \\ \alpha' &= D \sqrt{|K|((d-1) \vee 2)/2}.\end{aligned}$$

2.2 New variational formulas

Theorem [C. & F. Y. Wang (1997)].

$$\lambda_1 \geq \sup_{f \in \mathcal{F}} \inf_{r \in (0, D)} \frac{4f(r)}{\int_0^r C(s)^{-1} ds \int_s^D C(u) f(u) du}$$

Two notations:

$$C(r) = \cosh^{d-1} \left[\frac{r}{2} \sqrt{\frac{-K}{d-1}} \right], \quad r \in (0, D).$$

$$\mathcal{F} = \{f \in C[0, D] : f > 0 \text{ on } (0, D)\}.$$

Classical variational formula:

$$\lambda_1 = \inf \left\{ \int_M \|\nabla f\|^2 : f \in C^1, \pi(f) = 0, \pi(f^2) = 1 \right\}$$

Goes back to Lord S. J. W. Rayleigh(1877) or
E. Fischer (1905). Gen. R. Courant (1924).

Elementary functions: $1, \sin(\alpha r), \cosh^{1-d}(\alpha r) \sin(\beta r),$
 $\alpha = D \sqrt{|K|/(d-1)}/2, \quad \beta = \frac{\pi}{2D}.$

Corollary 1 [C. & F. Y. Wang (1997)].

$$\lambda_1 \geq \frac{dK}{d-1} \left\{ 1 - \cos^d \left[\frac{D}{2} \sqrt{\frac{K}{d-1}} \right] \right\}^{-1}, \quad d > 1, K \geq 0.$$

$$\lambda_1 \geq \frac{\pi^2}{D^2} \sqrt{1 - \frac{2D^2K}{\pi^4}} \cosh^{1-d} \left[\frac{D}{2} \sqrt{\frac{-K}{d-1}} \right],$$

$$d > 1, \quad K \leq 0.$$

Corollary 2 [C., E. Scacciatielli and L. Yao (2002)].

$$\lambda_1 \geq \frac{\pi^2}{D^2} + \frac{K}{2}, \quad K \in \mathbb{R}.$$

Representative test function:

$$f(r) = \left(\int_0^r C(s)^{-1} ds \right)^\gamma, \quad \gamma = \frac{1}{2}, 1.$$

$$C(s) = \cosh^{d-1} \left[\frac{s}{2} \sqrt{\frac{-K}{d-1}} \right].$$

$$\delta = \sup_{r \in (0, D)} \left(\int_0^r C(s)^{-1} ds \right) \left(\int_r^D C(s) ds \right).$$

Corollary 3 [C. (2000)]. $\lambda_1 \geq \xi_1$.

$$4\delta^{-1} \geq (\delta'_n)^{-1} \geq \xi_1 \geq \delta_n^{-1} \geq \delta^{-1},$$

$$\text{Explicit } (\delta'_n)^{-1} \downarrow, \quad \delta_n^{-1} \uparrow.$$

Compact M with convex ∂M . New, Lichnerowicz.

2.3 “Proof”: Estimation of λ_1

Step 1. g : eigenfunction: $Lg = -\lambda_1 g$, $g \neq \text{const.}$

From semigroups theory,

$$\frac{d}{dt} T_t g(x) = T_t Lg(x) = -\lambda_1 T_t g(x).$$

ODE. $T_t g(x) = g(x)e^{-\lambda_1 t}$. (1)

Step 2. Compact space. g Lipschitz w.r.t. ρ : c_g .

Key condition: $\tilde{T}_t \rho(x, y) \leq \rho(x, y) e^{-\alpha t}$. (2)

$$\begin{aligned} e^{-\lambda_1 t} |g(x) - g(y)| &\leq \tilde{T}_t |g(x) - g(y)| \quad (\text{by (1)}) \\ &\leq c_g \tilde{T}_t \rho(x, y) \\ &\leq c_g \rho(x, y) e^{-\alpha t} \quad (\text{by (2)}) \end{aligned}$$

for all t . Hence $\lambda_1 \geq \alpha$. **General!**

$$(2) \iff \tilde{L} \rho(x, y) \leq -\alpha \rho(x, y).$$

Two key points

- “Good” coupling: Classification of couplings. ρ -optimal couplings. “optimal mass transportation”.
- “Good” distance: Ord. \Rightarrow none. Reduce to dim. one.

$$L = a(x) \frac{d^2}{dx^2} + b(x) \frac{d}{dx}, \quad C(x) = \int_0^x \frac{b}{a}$$
$$f(x) \in C[0, D), \quad f|_{(0, D)} > 0$$
$$g(x) = \int_0^x e^{-C(y)} dy \int_y^D \frac{fe^C}{a} \quad \uparrow$$
$$\rho_f(x, y) = |g(x) - g(y)|.$$

2.4 Triangle matrix (birth-death process)

Notation:

$$\mu_0 = 1, \quad \mu_i = \frac{b_0 b_1 \cdots b_{i-1}}{a_1 a_2 \cdots a_i}, \quad i \geq 1.$$

$$Z = \sum_i \mu_i < \infty, \quad \pi_i = \frac{\mu_i}{Z}.$$

Two notations :

$\mathcal{F} = \{f : f_0 = 0, f \text{ is strictly increasing}\},$
 $\mathcal{F}' = \text{A slight modification of } \mathcal{F}.$

Theorem [C. (1996, 2000, 2001)].

- **Dual variational formulas.** Write $\bar{f} = f - \pi(f).$
$$\lambda_1 = \sup_{f \in \mathcal{F}} \inf_{i \geq 0} \mu_i b_i (f_{i+1} - f_i) / \sum_{j \geq i+1} \mu_j \bar{f}_j.$$

$$\lambda_1 = \inf_{f \in \mathcal{F}'} \sup_{i \geq 1} \mu_i b_i (f_{i+1} - f_i) / \sum_{j \geq i+1} \mu_j \bar{f}_j.$$
- **Explicit estimates.** $Z\delta^{-1} \geq \lambda_1 \geq (4\delta)^{-1}$, where
$$\delta = \sup_{i \geq 1} \sum_{j \leq i-1} (\mu_j b_j)^{-1} \sum_{j \geq i} \mu_j.$$
- **Approximation procedure.** \exists explicit η'_n, η''_n such that
$$\eta'_n{}^{-1} \geq \lambda_1 \geq \eta''_n{}^{-1} \geq (4\delta)^{-1}.$$

III. Basic inequalities and new forms of Cheeger's constants

3.1 Basic inequalities

(E, \mathcal{E}, π) : prob. space, $L^p(\pi)$, $\|\cdot\|_p$, $\|\cdot\| = \|\cdot\|_2$.
Dirichlet form $(D, \mathcal{D}(D))$ on $L^2(\pi)$.

Poincaré inequality : $\text{Var}(f) \leq CD(f)$, $C = \lambda_1^{-1}$.

Nash inequality : $\text{Var}(f) \leq CD(f)^{1/p} \|f\|_1^{2/q}$

J. Nash (1958) $1/p + 1/q = 1$

Logarithmic Sobolev inequality (L. Gross, 1976) :

$$\int_E f^2 \log \left(f^2 / \|f\|^2 \right) d\pi \leq CD(f).$$

3.2 Integral operator.

Symmetric form:

$$D(f) = \frac{1}{2} \int_{E \times E} J(\mathrm{d}x, \mathrm{d}y) [f(y) - f(x)]^2,$$

$$\mathcal{D}(D) = \{f \in L^2(\pi) : D(f) < \infty\},$$

$J \geq 0$: symmetric, no charge on $\{(x, x) : x \in E\}$.

$$\lambda_1 := \inf\{D(f) : \pi(f) = 0, \|f\| = 1\}.$$

Theorem [G. F. Lawler & A. D. Sokal (1988)] :

$$J(\mathrm{d}x, E)/\mathrm{d}\pi \leqslant M, \quad \lambda_1 \geqslant \frac{k^2}{2M}.$$

Six books:

- Chen, M. F. (1992), From Markov Chains to Non-Equilibrium Particle Systems, World Scientific, Singapore
- Sinclair, A. (1993), Algorithms for Random Generation and Counting: A Markov Chain Approach, Birkhäuser
- Colin de Verdière, Y. (1998), Spectres de Graphes, Publ. Soc. Math. France
- Chung, F. R. K. (1997), Spectral Graph Theory, CBMS, 92, AMS, Providence, Rhode Island L. (1997),

- Saloff-Coste, L. (1997), Lectures on finite Markov chains, LNM 1665, 301–413, Springer-Verlag
- Aldous, D. G. & Fill, J. A. (1994–), Reversible Markov Chains and Random Walks on Graphs

3.3 New results.

$$J^{(\alpha)}(dx, dy) = I_{\{r(x,y)^\alpha > 0\}} \frac{J(dx, dy)}{r(x, y)^\alpha}$$

$r \geq 0$, symmetric. $\alpha \in [0, 1]$

$J^{(1)}(dx, E)/\pi(dx) \leq 1$, π -a.e.

New forms of Cheeger's constants

Inequalities	Constant $k^{(\alpha)}$
Poincaré (C.&W.)	$\inf_{\pi(A) \in (0,1)} \frac{J^{(\alpha)}(A \times A^c)}{\pi(A) \wedge \pi(A^c)}$
Nash (C.)	$\inf_{\pi(A) \in (0,1)} \frac{J^{(\alpha)}(A \times A^c)}{[\pi(A) \wedge \pi(A^c)]^{(2q-3)/(2q-2)}}$
LogS (Wang) (C.)	$\lim_{r \rightarrow 0} \inf_{\pi(A) \in (0,r]} \frac{J^{(\alpha)}(A \times A^c)}{\pi(A) \sqrt{\log[e + \pi(A)^{-1}]}}$ $\lim_{\delta \rightarrow \infty} \inf_{\pi(A) > 0} \frac{J^{(\alpha)}(A \times A^c) + \delta \pi(A)}{\pi(A) \sqrt{1 - \log \pi(A)}}$

Theorem.

$k^{(1/2)} > 0 \implies$ corresponding inequality holds.

Four papers:

C. and Wang (1998),

C. (1999, 2000),

Wang (2001).

Estimates $\lambda_1 \geq \frac{k^{(1/2)2}}{1 + \sqrt{1 - k^{(1)2}}}.$ Can be sharp!

IV. New picture of ergodic theory and explicit criteria

4.1 Importance of the inequalities

$(D, \mathcal{D}(D)) \longrightarrow$ semigroup (P_t) : $P_t = e^{tL}$

Theorem [T.M.Liggett(89), L.Gross(76), C.(99)]

- Poincaré ineq. $\iff \text{Var}(P_t f) \leq \text{Var}(f) e^{-2\lambda_1 t}$.
- LogS \implies exponential convergence in entropy:
 $\text{Ent}(P_t f) \leq \text{Ent}(f) e^{-2\sigma t}$,
where $\text{Ent}(f) = \pi(f \log f) - \pi(f) \log \|f\|_1$.
- Nash ineq. $\iff \text{Var}(P_t f) \leq C \|f\|_1^2 / t^{q-1}$.

Nash inequality weakest?

4.2 Three traditional types of ergodicity.

$$\|\mu - \nu\|_{\text{Var}} = 2 \sup_A |\mu(A) - \nu(A)|$$

Ordinary erg. : $\lim_{t \rightarrow \infty} \|P_t(x, \cdot) - \pi\|_{\text{Var}} = 0$

Exp. erg. : $\lim_{t \rightarrow \infty} e^{\alpha t} \|P_t(x, \cdot) - \pi\|_{\text{Var}} = 0$

Strong erg. : $\lim_{t \rightarrow \infty} \sup_x \|P_t(x, \cdot) - \pi\|_{\text{Var}} = 0$

$$\iff \lim_{t \rightarrow \infty} e^{\beta t} \sup_x \|P_t(x, \cdot) - \pi\|_{\text{Var}} = 0$$

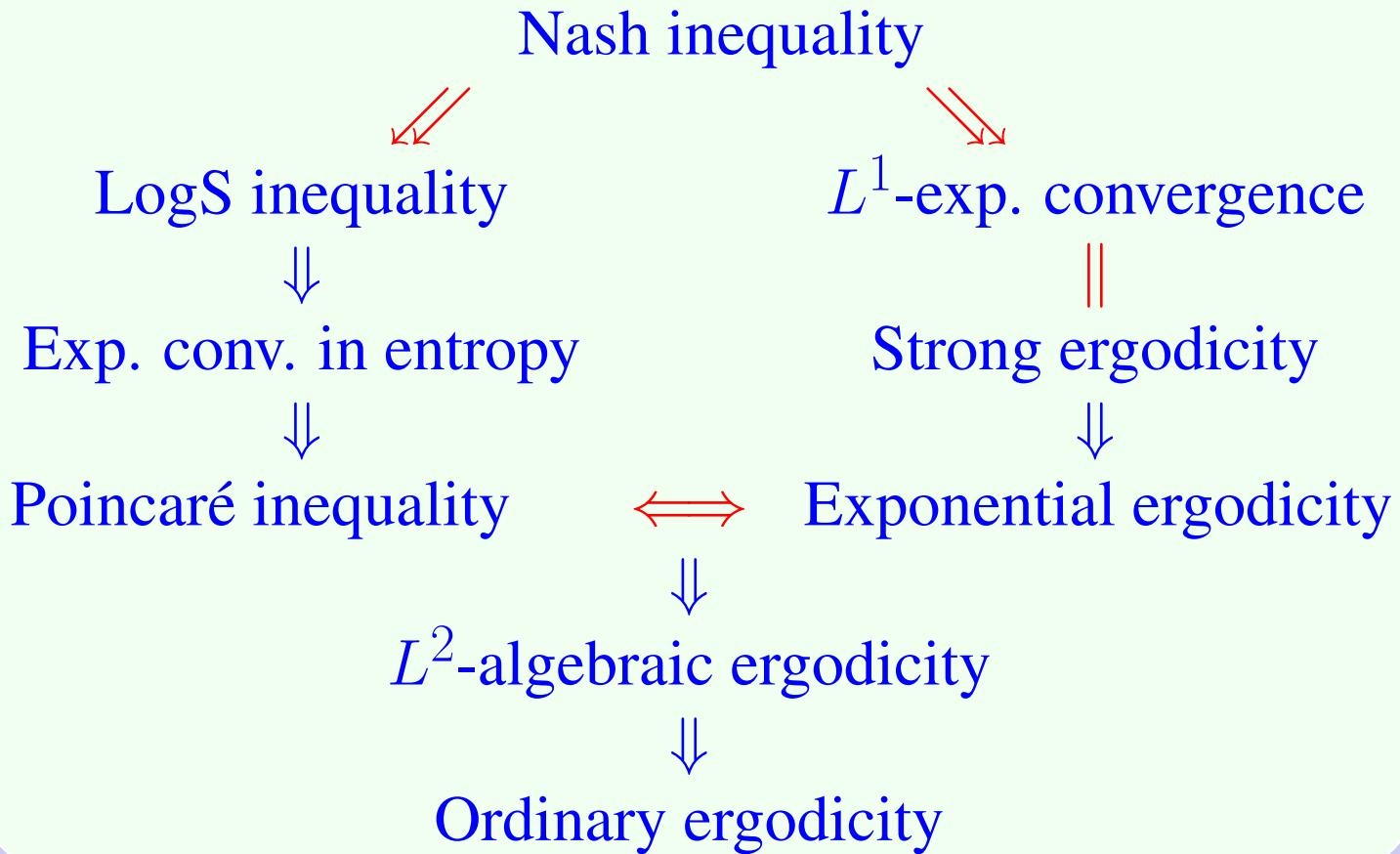
Classical Theorem :

Strong ergodicity \implies Exp. erg. \implies Ordinary erg.

LogS \implies Exp. conv. in entropy \implies Poincaré

4.3 New picture of ergodic theory

Theorem. For rever. Markov proc. with densities,



4.4 Explicit criteria for several types of ergodicity

Birth-death processes

$$\mu_0 = 1, \quad \mu_n = \frac{b_0 \cdots b_{n-1}}{a_1 \cdots a_n}, \quad n \geq 1;$$
$$\mu[i, k] = \sum_{i \leq j \leq k} \mu_j.$$

Theorem. Ten criteria for birth-death processes are listed in the following table [C. (2001)].

Property	Criterion
Uniq.	$\sum_{n \geq 0} \frac{1}{\mu_n b_n} \mu[0, n] = \infty \quad (*)$
Recur.	$\sum_{n \geq 0} \frac{1}{\mu_n b_n} = \infty$
Erg.	$(*) \text{ & } \mu[0, \infty) < \infty$
Exp. erg. Poincaré	$(*) \& \sup_{n \geq 1} \mu[n, \infty) \sum_{j \leq n-1} \frac{1}{\mu_j b_j} < \infty$
Dis. sp.	$(*) \& \lim_{n \rightarrow \infty} \sup_{k \geq n+1} \mu[k, \infty) \sum_{j=n}^{k-1} \frac{1}{\mu_j b_j} = 0$
LogS	$(*) \& \sup_{n \geq 1} \mu[n, \infty) \log[\mu[n, \infty)^{-1}] \sum_{j \leq n-1} \frac{1}{\mu_j b_j} < \infty$
Str. erg. L^1 -exp.	$(*) \& \sum_{n \geq 0} \frac{1}{\mu_n b_n} \mu[n+1, \infty) = \sum_{n \geq 1} \mu_n \sum_{j \leq n-1} \frac{1}{\mu_j b_j} < \infty$
Nash	$(*) \& \sup_{n \geq 1} \mu[n, \infty)^{(q-2)/(q-1)} \sum_{j \leq n-1} \frac{1}{\mu_j b_j} < \infty(\varepsilon)$

Contributors:

- Bobkov, S. G. and Götze, F. (1999a, b)
- C. (1991, 1996, 2000, 2001)
- Miclo, L. (1999, 2000)
- Mao, Y. H. (2001, 2002a, b)
- Wang, F. Y. (2001)
- Zhang, H. Z., Lin, X. and Hou, Z. T. (2000)
- Zhang, Y. H. (2001)
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<http://www.bnu.edu.cn/~chenmf>

The end!

Thank you, everybody!