

# Fixed-Point Iterations for Nonexpansive Maps

## The sharp bound of asymptotic regularity is $1/\sqrt{\pi}$

Roberto Cominetti

UNIVERSIDAD DE CHILE

[rccc@dii.uchile.cl](mailto:rccc@dii.uchile.cl)

Based on joint work with

**J-B. Baillon, M. Bravo, J. Soto, J. Vaisman**

CELEBRATING TERRY ROCKAFELLAR'S 80TH BIRTHDAY

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# Banach-Picard fixed point iteration

$T : X \rightarrow X$  contraction

(BP)

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$$\|x^{n+1} - x^n\| = \|Tx^n - x^n\| \leq \rho^n \|Tx^0 - x^0\| \rightarrow 0$$



convergence + error estimates + stopping rule

# Krasnoselskii-Mann fixed point iteration

$T : C \rightarrow C$  non-expansive /  $C$  convex bounded in  $(X, \|\cdot\|)$

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**Question:**  $\|Tx^n - x^n\| \rightarrow 0$  ? (Asymptotic Regularity)

# History of $\|Tx^n - x^n\| \rightarrow 0$

- 1955 **Krasnoselskii**:  $X$  uniformly convex,  $C$  compact,  $\alpha_n \equiv 1/2$
- 1957 **Schaefer**: extends Krasnoselskii to  $\alpha_n \equiv \alpha$
- 1966 **Ishikawa**: extends Krasnoselskii to  $X$  strictly convex
- 1966 **Browder-Petryshyn**:  $X$  uniformly convex,  $\text{Fix}(T) \neq \emptyset$ ,  $\alpha_n \equiv \alpha$
- 1976 **Ishikawa**:  $X$  general,  $C$  compact,  $\sum \alpha_n = \infty$ ,  $\alpha_n \leq 1 - \epsilon$
- 1978 **Edelstein-O'Brien**:  $\|Tx^n - x^n\| \rightarrow 0$  uniformly w.r.t  $x_0$
- 1983 **Goebel-Kirk**: uniformly w.r.t.  $x_0$  and  $T$  (for  $C$  given)
- 1992 **Baillon-Bruck**:  $\alpha_n \equiv \alpha \Rightarrow \|Tx^n - x^n\| = O(1/\log n)$
- 2003 **Kohlenbach**:  $\|Tx^n - x^n\| \rightarrow 0$  only depends on  $\text{diam}(C)$

# Baillon-Bruck's conjecture (1992)

There exists a universal constant  $\kappa$  such that

$$(BB) \quad \|Tx^n - x^n\| \leq \kappa \frac{\text{diam}(C)}{\sqrt{\sum_{k=1}^n \alpha_k(1-\alpha_k)}}$$

REMARK: in continuous time  $\|Tx(t) - x(t)\| \leq \kappa \frac{\text{diam}(C)}{\sqrt{t}}$

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When  $\alpha_n \equiv \alpha$  the bound holds with  $\kappa = 1/\sqrt{\pi}$ .

Our contributions:

- The bound holds for general  $\alpha_n$  with  $\kappa = 1/\sqrt{\pi} \sim 0.5642$
- Nonlinear maps: the constant  $\kappa = 1/\sqrt{\pi}$  is optimal
- Affine maps: sharp bound with  $\kappa = \max_z \sqrt{z} e^{-z} I_0(z) \sim 0.4688$

# Nonlinear Maps

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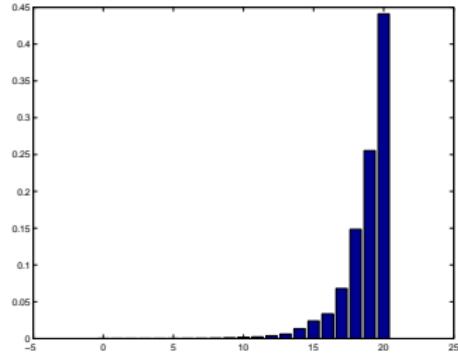
Recall

$$x^{n+1} = (1 - \alpha_{n+1})x^n + \alpha_{n+1}Tx^n$$

# Alternative expression for $x^n$

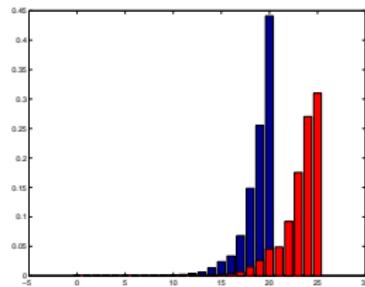
Let  $\pi_i^n = \alpha_i \prod_{i+1}^n (1 - \alpha_k)$  and set  $Tx^{-1} = x_0$  by convention, then

$$x^n = \sum_{i=0}^n \pi_i^n Tx^{i-1}$$



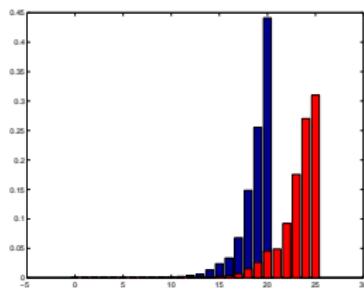
# A recursive bound $\|x^m - x^n\| \leq d_{mn}$

$$x^m - x^n = \sum_{j=0}^m \pi_j^m T x^{j-1} - \sum_{i=0}^n \pi_i^n T x^{i-1}$$



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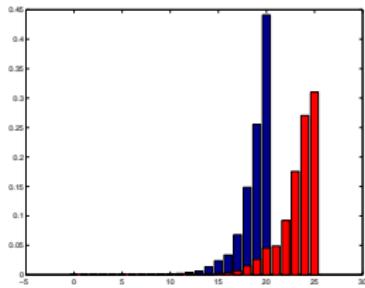


Let  $P_{mn}$  be the set of transport plans  $z \geq 0$  taking  $\pi^n$  to  $\pi^m$

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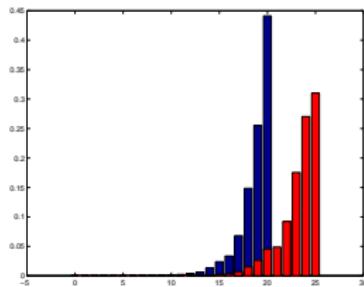


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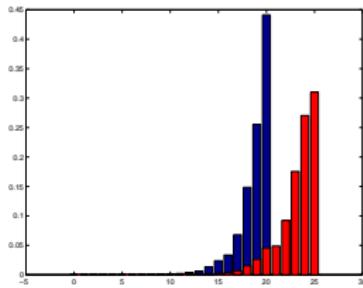


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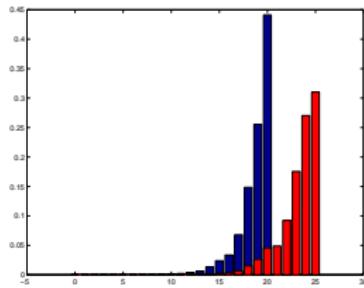


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$$\|x^m - x^n\| \leq \sum_{j=0}^m \sum_{i=0}^n z_{ji} d_{j-1, i-1} \quad \leftarrow \quad \min_z$$



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Set  $d_{-1,n} = 1$ ,  $d_{-1,-1} = 0$ , and define inductively

$$(R) \quad d_{mn} \triangleq \min_{z \in P_{mn}} \sum_{j=0}^m \sum_{i=0}^n z_{ji} d_{j-1,i-1}$$

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Theorem (Bravo-C. – Sep. 2014)

*Bound (R) is best possible: There exists a non-expansive  $T$  on the set  $C = [0, 1]^{\mathbb{N}} \subseteq \ell^\infty(\mathbb{N})$  which attains  $\|x^m - x^n\| = d_{mn}$  for all  $m, n$ .*

Proof: Built from dual solutions of the optimal transports.

# Restatement of (BB)

$$\|Tx^n - x^n\| = \left\| \frac{x^{n+1} - x^n}{\alpha_{n+1}} \right\| \leq \frac{d_{n,n+1}}{\alpha_{n+1}} = ?$$

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↓

$$\frac{d_{n,n+1}}{\alpha_{n+1}} \leq \frac{1}{\sqrt{\pi}} \frac{1}{\sqrt{\sum_{k=1}^n \alpha_k(1-\alpha_k)}} \quad ?$$

# Upper estimate: $d_{mn} \leq c_{mn}$

Consider the non-optimal transport plan

$$z_{ji} = \begin{cases} \pi_i^n & \text{for } j = i \leq m \\ \pi_j^m \pi_i^n & \text{for } i = m+1, \dots, n \\ 0 & \text{otherwise} \end{cases}$$

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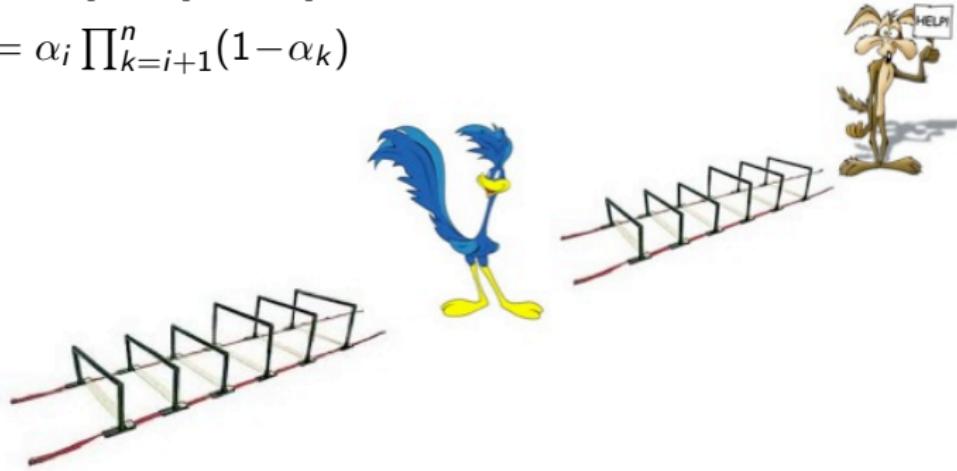
Setting  $c_{-1,n} = 1$ ,  $c_{-1,-1} = 0$ , we get inductively

$$\|x^m - x^n\| \leq d_{mn} \leq c_{mn} \triangleq \sum_{j=0}^m \sum_{i=m+1}^n \pi_j^m \pi_i^n c_{j-1,i-1}$$

# Probabilistic interpretation of the recursion

$$\mathbb{P}[C_i = 1] = \mathbb{P}[R_i = 1] = \alpha_i$$

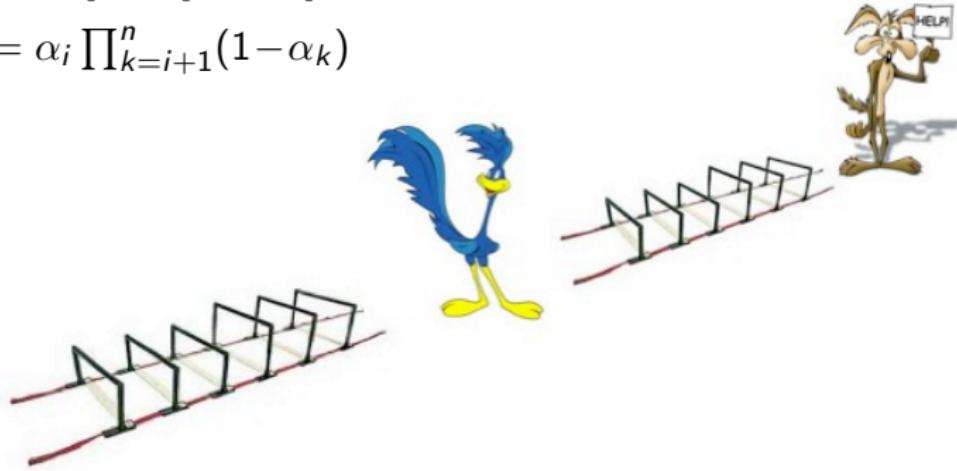
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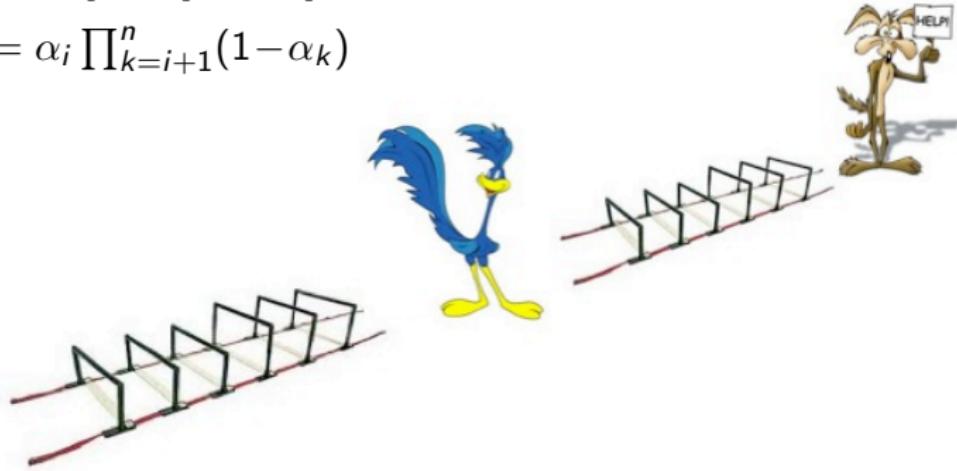


$$c_{mn} = \mathbb{P}[\text{roadrunner escapes}]$$

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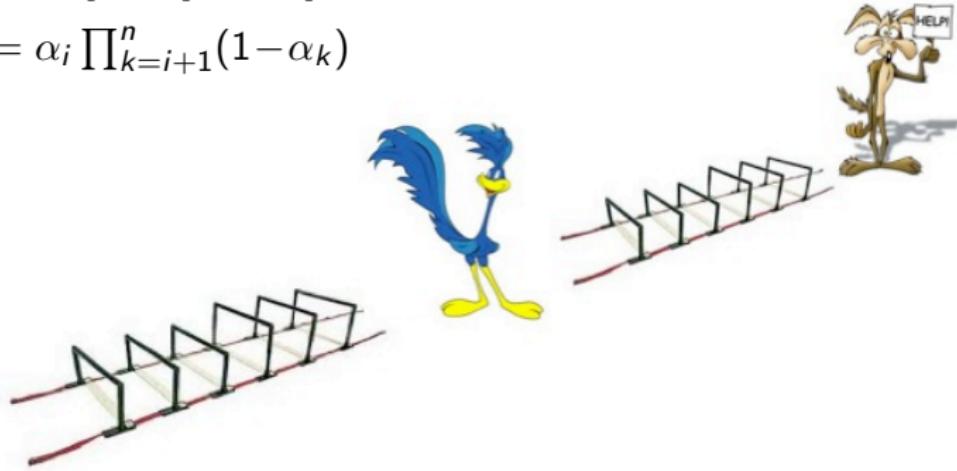


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$$c_{mn} = \mathbb{P}[\sum_k^n C_i > \sum_k^m R_i, \forall k = m+1, \dots, 1]$$

Coyote must fall more often than Roadrunner

# The random walk and the gambler's ruin appear...

$$c_{n,n+1} = \mathbb{P}[\sum_k^{n+1} C_i > \sum_k^n R_i, \forall k = n+1, \dots, 1]$$

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$$Z_i = C_i - R_i = \begin{cases} -1 & \text{pbb} & \alpha_i(1 - \alpha_i) \\ 0 & \text{pbb} & 1 - 2\alpha_i(1 - \alpha_i) \\ 1 & \text{pbb} & \alpha_i(1 - \alpha_i) \end{cases}$$

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⇒ random walk on  $\mathbb{Z}$  that moves with probability  $p_i = 2\alpha_i(1 - \alpha_i)$  and then tosses a coin to decide whether to go left or right

$$\|Tx^n - x^n\| \leq \frac{c_{n,n+1}}{\alpha_{n+1}} = \mathbb{P}[\text{process } \geq 0 \text{ over } n \text{ stages}]$$

# An explicit formula for the bound

Rewrite  $Z_i = M_i D_i$  with  $M_i = \text{move}/\text{stay}$  and  $D_i = \text{direction}$

$$M_i = \begin{cases} 1 & \text{pb} \\ 0 & \text{pb} \end{cases} \quad \begin{matrix} p_i \\ 1 - p_i \end{matrix} ; \quad D_i = \begin{cases} -1 & \text{pb} \\ 1 & \text{pb} \end{cases} \quad \begin{matrix} \frac{1}{2} \\ \frac{1}{2} \end{matrix}$$

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Conditional on the number of moves  $M = M_1 + \dots + M_n = m$ , this is a standard random walk on  $m$  stages. The probability for the latter to remain non-negative is  $F(m) = \binom{m}{\lfloor m/2 \rfloor} 2^{-m}$ , therefore

$$\|x_n - Tx_n\| \leq \frac{c_{n,n+1}}{\alpha_{n+1}} = \sum_{m=0}^n F(m) \mathbb{P}[M = m] = \mathbb{E}[F(M)]$$

# Sharp bound

Thus  $(BB)$  has been reduced to

$$\mathbb{E}[F(M)] \leq \frac{1}{\sqrt{\pi \sum_{i=1}^n \alpha_i(1 - \alpha_i)}}$$

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$$\sqrt{\frac{\pi}{2}(p_1 + \dots + p_n)} \mathbb{E}[F(M_1 + \dots + M_n)] \leq 1$$

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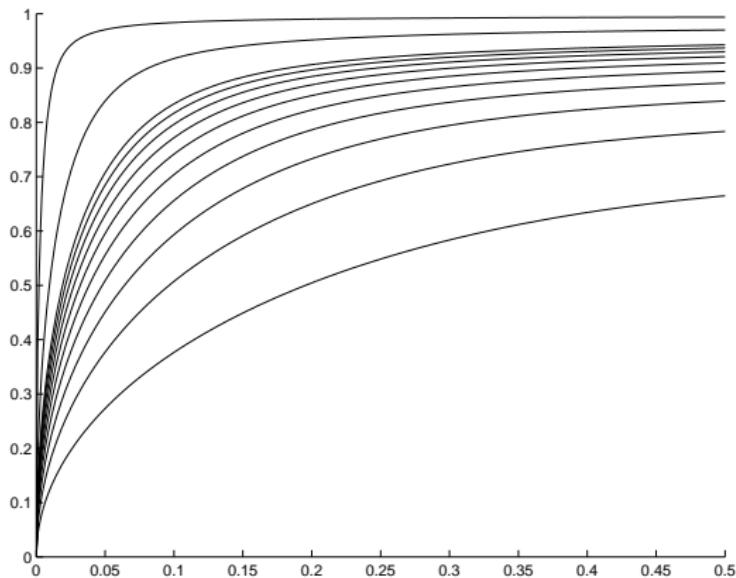
$$\underbrace{\sqrt{\frac{\pi}{2}(p_1 + \dots + p_n)} \mathbb{E}[F(M_1 + \dots + M_n)]}_{R(p)} \leq 1$$

## Lemma

$R(p)$  is maximal when  $p_i \in \{u, \frac{1}{2}\}$  for some  $0 < u < \frac{1}{2}$

# Sharp bound: all $p_i = u$

$$R(p) = \sqrt{\frac{\pi}{2} n u} \mathbb{E}[F(B(n, u))] = \sqrt{\frac{\pi}{2} n u} {}_2F_1(-n, \frac{1}{2}; 2; 2u)$$



# Sharp bound: some $p_i = \frac{1}{2}$

Suppose  $p_1 = \frac{1}{2}$  and let  $S = M_2 + \dots + M_n$ . Conditioning on  $M_1$

$$\mathbb{E}[F(M)] = \mathbb{E}[G(S)]$$

where  $G(k) = \frac{1}{2}[F(k) + F(k+1)]$ .

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This  $G$  is convex so we may use the following Hoeffding-type inequality

Theorem (C-Soto-Vaisman'2014)

Let  $Z$  be Poisson with  $z = \mathbb{E}(Z) = \mathbb{E}(S)$ . Then  $\mathbb{E}[G(S)] \leq \mathbb{E}[G(Z)]$ .

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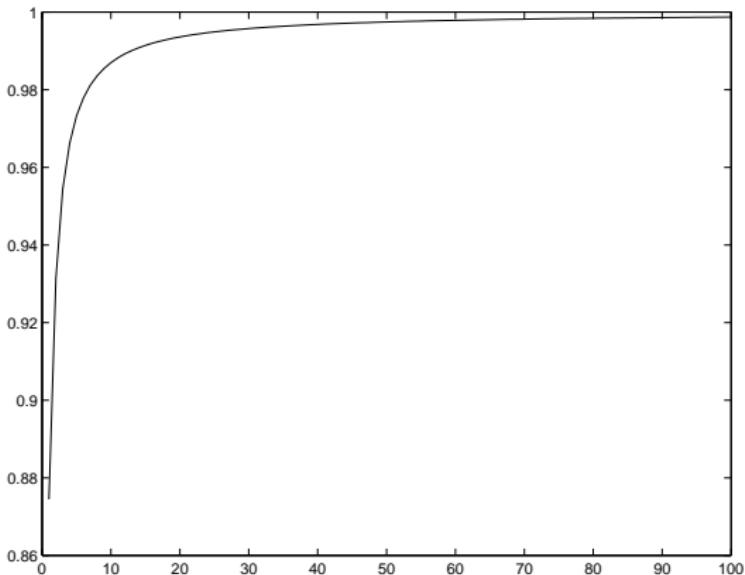
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$$\Rightarrow \mathbb{E}[F(M)] \leq \mathbb{E}[G(Z)] = I_0(z) + (1 - \frac{1}{2z})I_1(z)$$

with  $I_0(z), I_1(z)$  modified Bessel functions

# Sharp explicit bound: some $p_i = \frac{1}{2}$

$$R(p) \leq \sqrt{\frac{\pi}{2} \left( \frac{1}{2} + z \right)} [l_0(z) + (1 - \frac{1}{2z}) l_1(z)]$$



# Asymptotic regularity for nonlinear maps

Combining the previous estimates we finally get

Theorem (C-Soto-Vaisman'2014)

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with  $\kappa = 1/\sqrt{\pi} \sim 0.5642$

# The constant $1/\sqrt{\pi}$ is optimal

Reconsider the optimal recursive bounds

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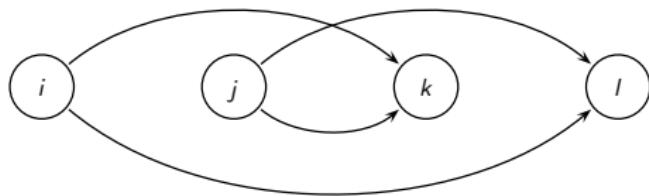
# The constant $1/\sqrt{\pi}$ is optimal

Reconsider the optimal recursive bounds

$$(R) \quad d_{mn} \triangleq \min_{z \in P_{mn}} \sum_{j=0}^m \sum_{i=0}^n z_{ji} d_{j-1,i-1}$$

Theorem (Aygen-Satik'2004)

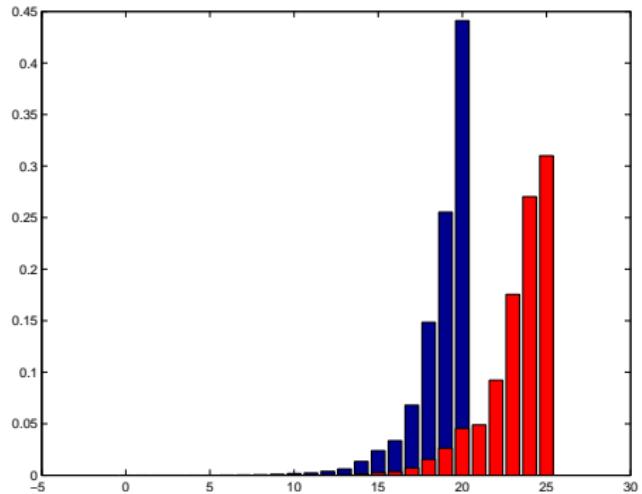
The recursion (R) defines a metric on the set  $\{-1, 0, 1, 2, 3, \dots\}$ . Moreover, for  $i < j < k < l$  we have  $d_{ik} + d_{jl} \geq d_{il} + d_{jk}$ .



Original proof 80+ pages. Short proof by Bravo-C. (Nov. 2014).

# No flow-crossing in optimal transports

$$d_{ik} + d_{jl} \geq d_{il} + d_{jk}$$



# No flow-crossing in optimal transports

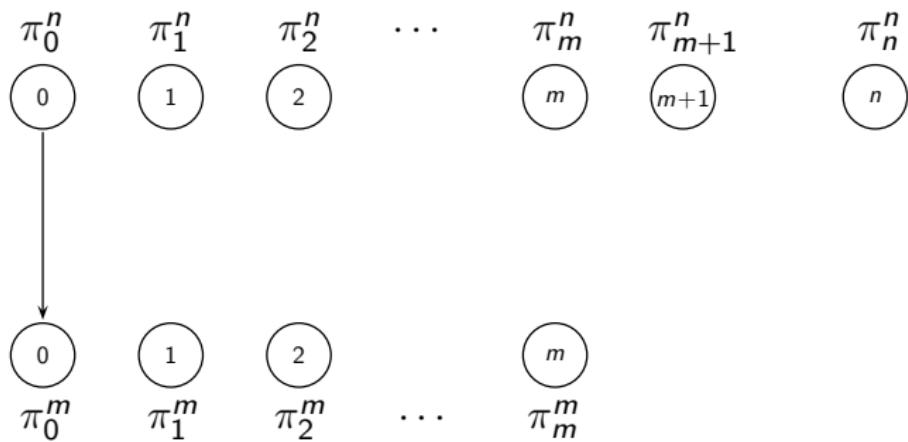
For  $\alpha_n \equiv \alpha \geq \frac{1}{2}$  the optimal transport is

$$\begin{array}{ccccccc} \pi_0^n & \pi_1^n & \pi_2^n & \cdots & \pi_m^n & \pi_{m+1}^n & \pi_n^n \\ (0) & (1) & (2) & & (m) & (m+1) & (n) \end{array}$$

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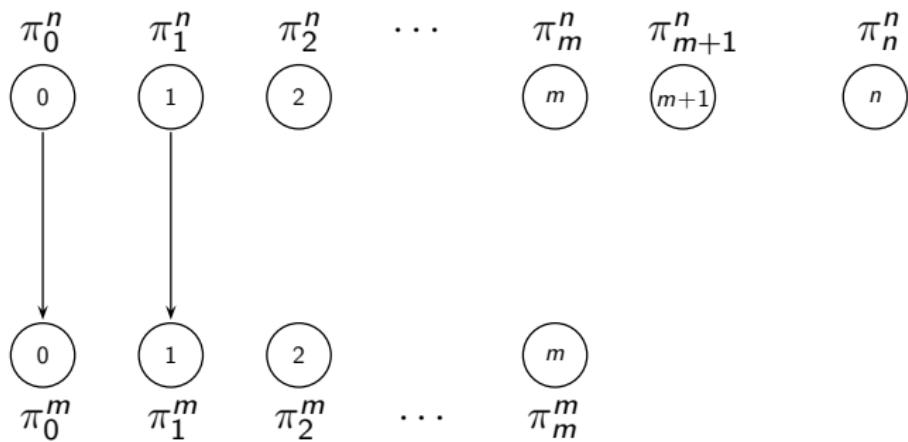
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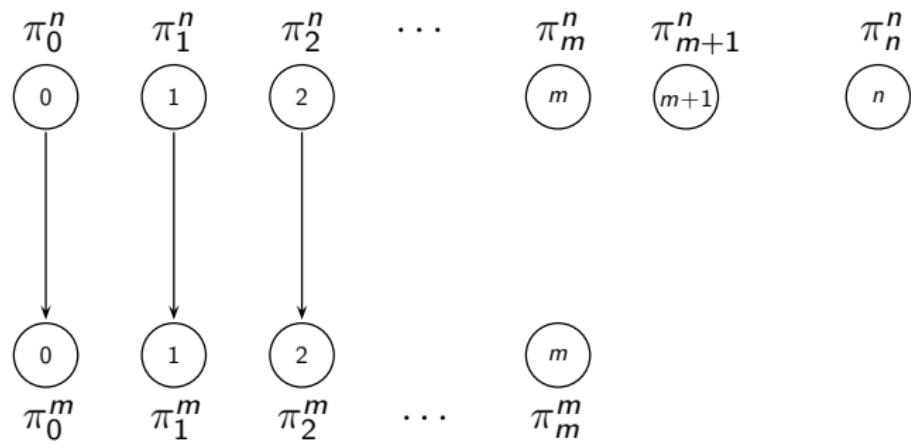
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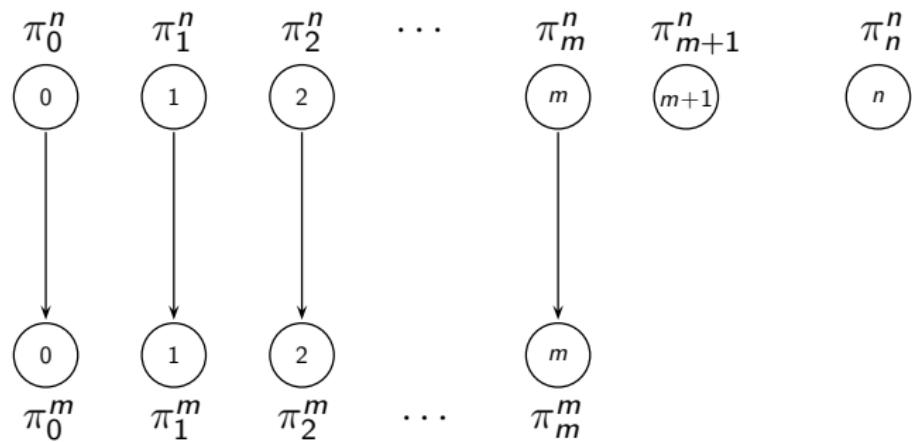
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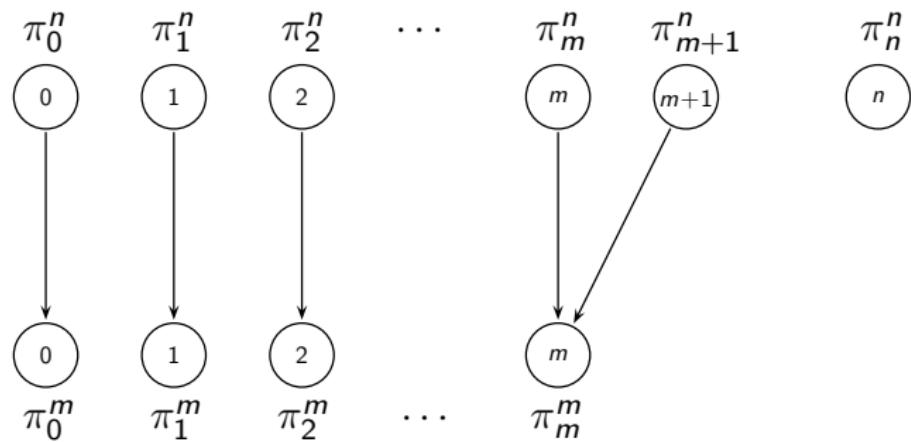
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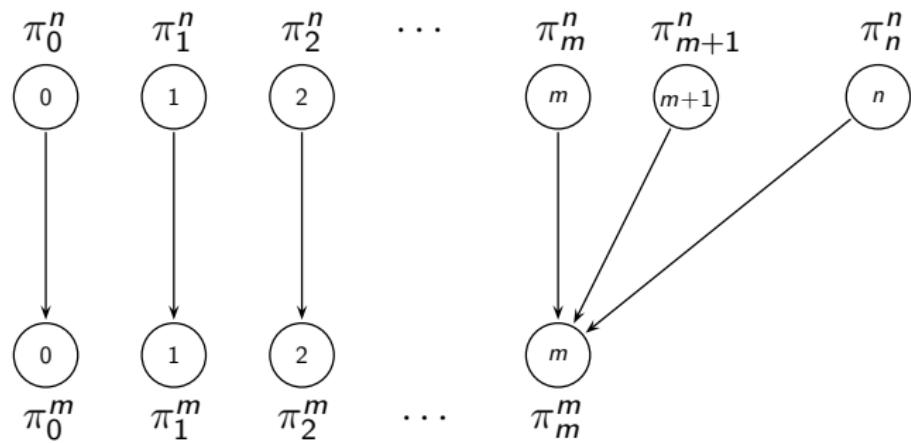
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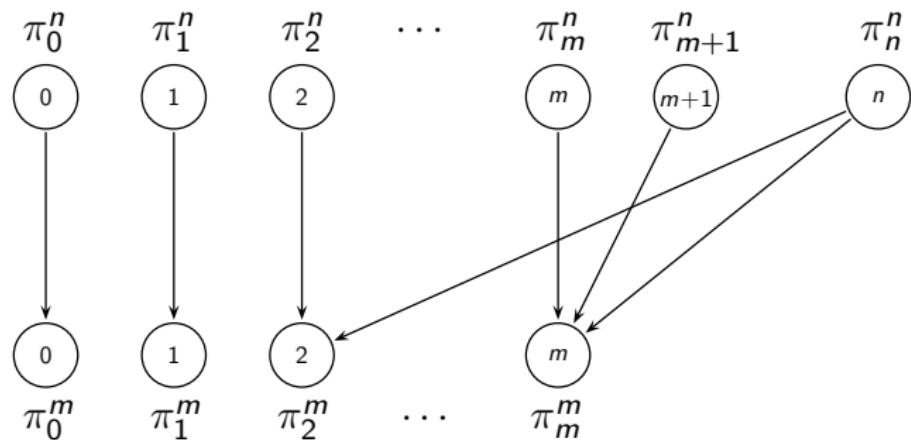
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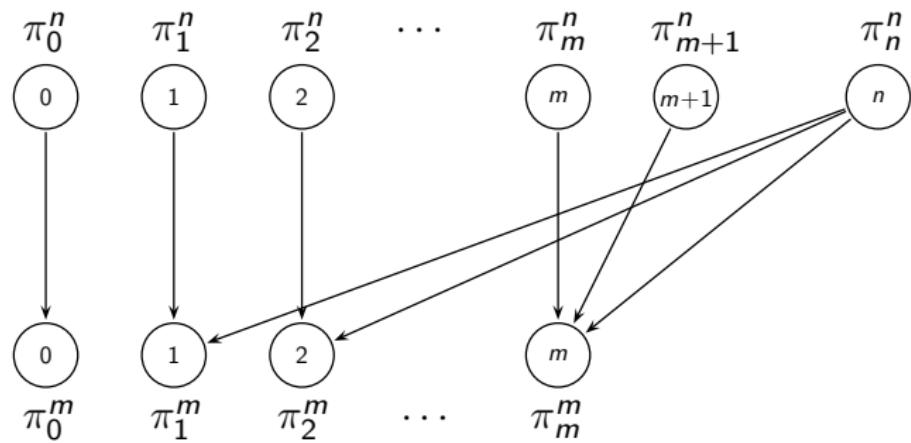
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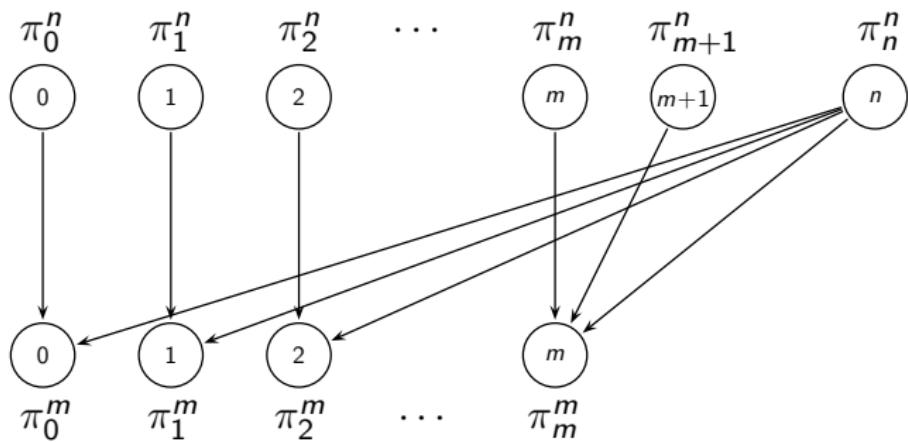
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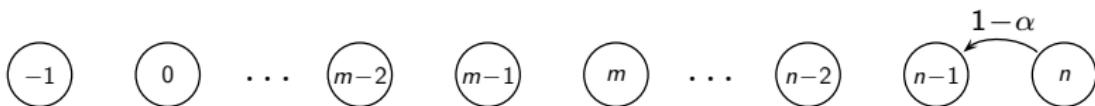
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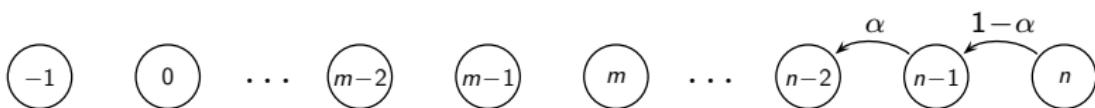
# Stochastic process for $d_{mn}$ (case $\alpha_n \equiv \alpha \geq \frac{1}{2}$ )



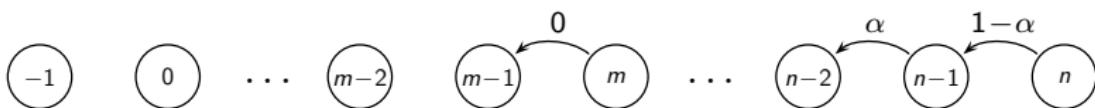
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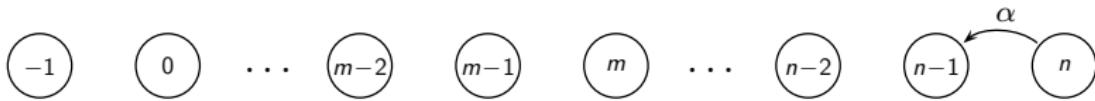
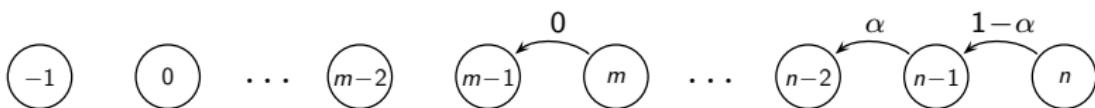
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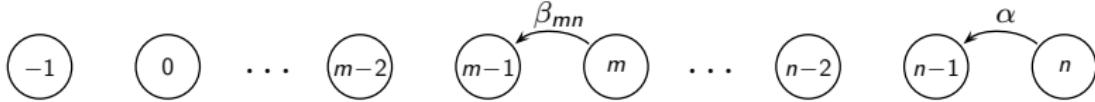
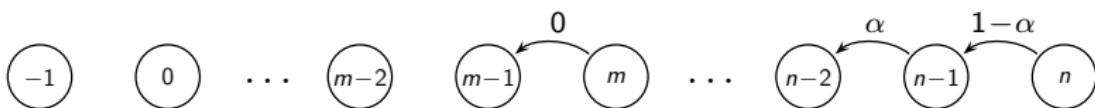
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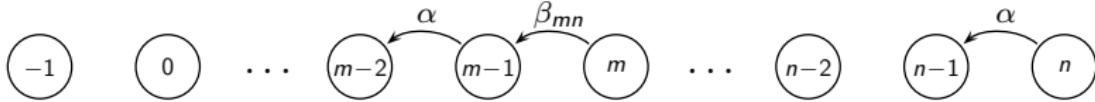
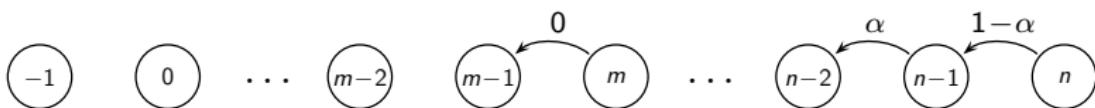
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# The constant $\frac{1}{\sqrt{\pi}}$ is optimal

Transition probabilities

Process  $d_{mn}$ :  $P_{mn}^{ij} = \mathbb{P}((m, n) \searrow (i, j))$

Process  $c_{mn}$ :  $Q_{mn}^{ij} = \mathbb{P}((m, n) \searrow (i, j))$

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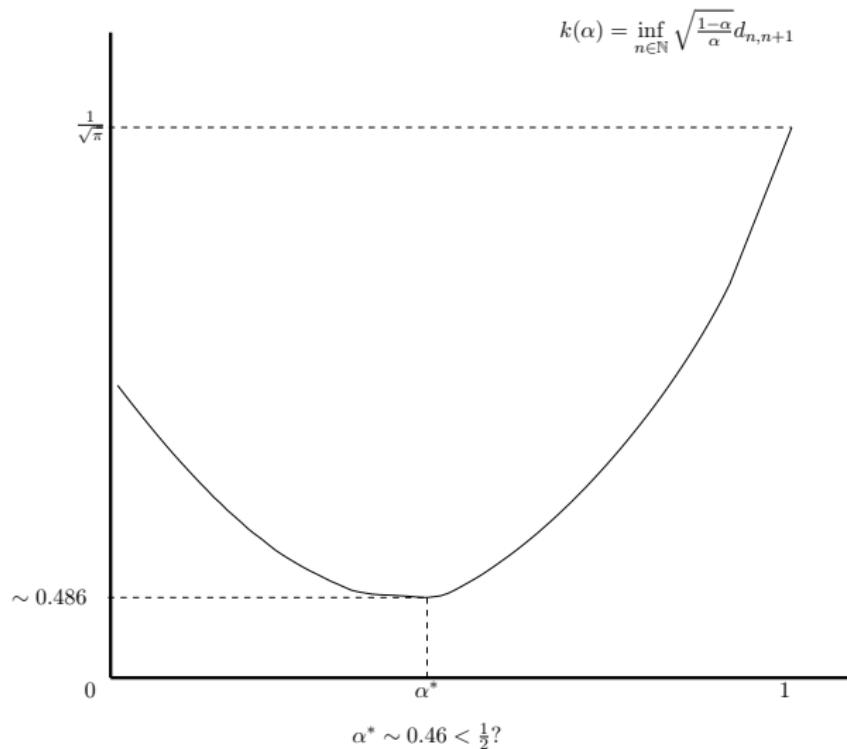
Taking  $\alpha_n = 1 - \log n/n$  we get

$$\lim_n \sqrt{n\alpha_n(1-\alpha_n)} \frac{d_{n,n+1}}{\alpha_n} = \lim_n \sqrt{n\alpha_n(1-\alpha_n)} \frac{c_{n,n+1}}{\alpha_n} = \frac{1}{\sqrt{\pi}}$$

Theorem (Bravo-C., April 2015)

*The constant  $\frac{1}{\sqrt{\pi}}$  in (BB) is sharp.*

# Optimal constant for fixed $\alpha$ ?



$$\|Tx^n - x^n\| \leq \frac{1}{\sqrt{\pi}} \frac{\text{diam}(C)}{\sqrt{\sum_{k=1}^n \alpha_k(1-\alpha_k)}}$$

**Feliz cumpleaños Terry !**

# Affine Maps

# Affine maps

**Example:** Right-shift on  $\ell^1(\mathbb{N})$

$C = \{p \in \ell^1(\mathbb{N}) : p_i \geq 0, \sum_{i=0}^{\infty} p_i = 1\}$  with  $\text{diam}(C) = 2$

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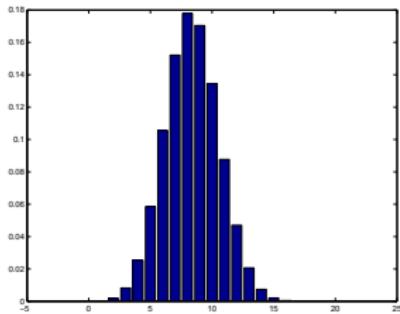
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$$x_k^n = \mathbb{P}(X_1 + \dots + X_n = k)$$

$$X_i \sim \text{Bernoulli}(\alpha_i)$$

$$\|Tx^n - x^n\|_1 = 2 \max_k x_k^n$$

# Sums of Bernoullis and (BB)

Theorem (Baillon-C-Vaisman, arXiv'2013)

Let  $X_i$  be independent Bernoullis with  $\mathbb{P}(X_i = 1) = \alpha_i$ . Then

$$p_k^n = \mathbb{P}(X_1 + \dots + X_n = k) \leq \frac{\eta}{\sqrt{\sum_{i=1}^n \alpha_i(1 - \alpha_i)}}$$

where  $\eta = \max_{u \geq 0} \sqrt{u} e^{-u} I_0(u) \sim 0.4688$  with  $I_0(\cdot)$  modified Bessel function. This bound is sharp.

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Corollary

For the right shift in  $\ell^1(\mathbb{N})$  the optimal bound in (BB) is  $\kappa = \eta$ .

# Asymptotic regularity for affine maps

Let  $\bar{x} \in \text{Fix } T$  and  $C = B(\bar{x}, r)$  with  $r = \|x^0 - \bar{x}\|$  so that  $T : C \rightarrow C$ .

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## Corollary

For affine maps (BB) holds with  $\kappa = \eta$ . This bound is sharp and is attained by the right shift.