

# Zero-one laws for uniform inhomogeneous Diophantine approximations

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# Introduction

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# Notation

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We will use the following conventions:

- $m$  and  $n$  are positive integers and  $d = m + n$ .  $\mathbf{M}_{n,m}$  is the set of all real  $n \times m$  matrices.

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$$\Theta = \begin{pmatrix} \theta_{1,1} & \cdots & \theta_{1,m} \\ \cdots & \cdots & \cdots \\ \theta_{n,1} & \cdots & \theta_{n,m} \end{pmatrix} \in \mathbf{M}_{n,m}, \quad \boldsymbol{\eta} = \begin{pmatrix} \eta_1 \\ \vdots \\ \eta_n \end{pmatrix} \in \mathbb{R}^n$$

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— a pair of a real  $n \times m$  matrix and  $n$ -vector.

- $|\cdot|$  is the supremum norm of a vector, and  $\|\cdot\|_{\mathbb{Z}^n}$  is the distance to the nearest integer vector function.

# Notation

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- Classical homogeneous Diophantine approximations studies the form

$$||\Theta \mathbf{q}||_{\mathbb{Z}^n}$$

- In the inhomogeneous setting, we replace the linear for with affine: we study

$$||\Theta \mathbf{q} - \boldsymbol{\eta}||_{\mathbb{Z}^n}$$

More specifically, we want to compare the values of this forms with values of certain functions:

# Uniform and asymptotic approximations

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Fix a non-increasing real function  $g$ . We study the system of inequalities

$$\begin{cases} \|\Theta \mathbf{q} - \boldsymbol{\eta}\|_{\mathbb{Z}^n}^n \leq g(T) \\ |\mathbf{q}|^m \leq T \end{cases} \quad (1)$$

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In both homogeneous and inhomogeneous setups, two types of questions are of interest:

- Uniform approximations: does the system (1) have an integer solution  $\mathbf{q}$  for all  $T$  large enough?  $\hat{\mathbf{D}}_{n,m}[g]$
- Asymptotic approximations: does the system (1) have an integer solution  $\mathbf{q}$  for an unbounded set of  $T$ ?  $\hat{\mathbf{W}}_{n,m}[g]$

# Main definitions

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## Definition ( $g$ -Dirichlet pairs)

By  $\widehat{\mathbf{D}}_{n,m}[g]$  we will denote the set of pairs  $(\Theta, \boldsymbol{\eta})$  for which the system

$$\begin{cases} \|\Theta \mathbf{q} - \boldsymbol{\eta}\|_{\mathbb{Z}^n}^n \leq g(T) \\ |\mathbf{q}|^m \leq T \end{cases}$$

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## Definition ( $g$ -approximable pairs)

By  $\widehat{\mathbf{W}}_{n,m}[g]$  we will denote the set of pairs  $(\Theta, \boldsymbol{\eta})$  for which the inequality

$$\|\Theta \mathbf{q} - \boldsymbol{\eta}\|_{\mathbb{Z}^n}^n \leq g(|\mathbf{q}|^m)$$

has infinitely many integer solutions  $\mathbf{q}$ . *Homogeneous case:*  $\mathbf{W}_{n,m}[g]$ .

# General weight functions

## Definition

Let

$$\underline{\alpha} = (\alpha_1, \dots, \alpha_n) \quad \text{and} \quad \underline{\beta} = (\beta_1, \dots, \beta_m)$$

be the  $n$ -tuple and  $m$ -tuple of positive, increasing, continuous real valued functions, defined on the interval  $[1, +\infty)$  and having infinite limit an infinity. It will be convenient to assume that  $\alpha_i(1) = \beta_j(1) = 1$ . One can extend these functions to the whole set of nonnegative real arguments by defining

$$\alpha_i(T) = \frac{1}{\alpha_i\left(\frac{1}{T}\right)} \quad \text{and} \quad \beta_j(T) = \frac{1}{\beta_j\left(\frac{1}{T}\right)}$$

for  $0 < T \leq 1$ , and, by continuity,  $\alpha_i(0) = \beta_j(0) = 0$ .

# General weight functions

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For  $\mathbf{y} = (y_1, \dots, y_n)$  and  $\mathbf{x} = (x_1, \dots, x_m)$ , we write

$$|\mathbf{y}|_{\underline{\alpha}} = \max_{1 \leq i \leq n} \alpha_i^{-1}(|y_i|) \quad \text{and} \quad |\mathbf{x}|_{\underline{\beta}} = \max_{1 \leq j \leq m} \beta_j^{-1}(|x_j|).$$

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By  $\|\cdot\|_{\underline{\alpha}}$  and  $\|\cdot\|_{\underline{\beta}}$  we will denote the corresponding "distance to the nearest integer" functions, that is,

$$\|\mathbf{y}\|_{\underline{\alpha}} = \min_{\mathbf{p} \in \mathbb{Z}^n} |\mathbf{y} - \mathbf{p}|_{\underline{\alpha}} \quad \text{and} \quad \|\mathbf{x}\|_{\underline{\beta}} = \min_{\mathbf{q} \in \mathbb{Z}^n} |\mathbf{x} - \mathbf{q}|_{\underline{\beta}}.$$

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## Example

Let  $\alpha_i(T) = \beta_j(T) = T$ . Then  $|\mathbf{y}|_{\underline{\alpha}} = |\mathbf{y}|$  and  $|\mathbf{x}|_{\underline{\beta}} = |\mathbf{x}|$ .

# Approximations with general weight functions

## Definition

1.  $\Theta^\top$  is  $(f, \underline{\beta}, \underline{\alpha})$ -approximable if the inequality

$$\|\Theta^\top \mathbf{y}\|_{\underline{\beta}} \leq f(|\mathbf{y}|_{\underline{\alpha}})$$

has infinitely many solutions  $\mathbf{y} \in \mathbb{Z}^n \setminus \{\mathbf{0}\}$ .

$$\mathbf{W}_{m,n}[f; \underline{\beta}, \underline{\alpha}]$$

2.  $(\Theta, \boldsymbol{\eta})$  is  $(g, \underline{\alpha}, \underline{\beta})$ -Dirichlet if

$$\begin{cases} \|\Theta \mathbf{q} - \boldsymbol{\eta}\|_{\underline{\alpha}} \leq g(T), \\ |\mathbf{q}|_{\underline{\beta}} \leq T \end{cases}$$

has a solution  $\mathbf{q} \in \mathbb{Z}^m \setminus \{\mathbf{0}\}$  for any  $T \in \mathbb{R}_+$  large enough.

$$\hat{\mathbf{D}}_{n,m}[g; \underline{\alpha}, \underline{\beta}]$$

# Approximations with general weight functions

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1.  $\mathbf{W}_{m,n}[f; \underline{\beta}, \underline{\alpha}]$ :

$$\begin{cases} \|\left(\Theta^\top\right)_j \mathbf{y}\| \leq \beta_j(f(T)), & j = 1, \dots, m; \\ |y_i| \leq \alpha_i(T), & i = 1, \dots, n \end{cases}$$

has a solution  $\mathbf{y} = (y_1, \dots, y_n) \in \mathbb{Z}^n$  for an unbounded set of  $T \in \mathbb{R}_+$ .

2.  $\widehat{\mathbf{D}}_{n,m}[g; \underline{\alpha}, \underline{\beta}]$ :

$$\begin{cases} \|\Theta_i \mathbf{q} - \eta_i\| \leq \alpha_i(g(T)), & i = 1, \dots, n; \\ |q_j| \leq \beta_j(T), & j = 1, \dots, m \end{cases}$$

has a solution  $\mathbf{q} = (q_1, \dots, q_m) \in \mathbb{Z}^m$  for any  $T \in \mathbb{R}_+$  large enough.

# Quasimultiplicativity

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An increasing function  $h : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  is *quasimultiplicative* if  $\exists M > 1 : \forall |k| \gg 1$

$$c_1 h(M^k) \leq h(M^{k+1}) \leq c_2 h(M^k), \quad (2)$$

where  $1 < c_1 \leq c_2$  are some absolute constants.

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- $h + r$ ,  $h \cdot r$ ,  $h \circ r$  and  $r \circ h$  are quasimultiplicative functions.
- $\frac{1}{h}$  and  $h^{-1}$  are quasimultiplicative.

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- $h + r$ ,  $h \cdot r$ ,  $h \circ r$  and  $r \circ h$  are quasimultiplicative functions.
- $\frac{1}{h}$  and  $h^{-1}$  are quasimultiplicative.
- if  $h$  is a quasimultiplicative function for which

$$h(T) = \frac{1}{h\left(\frac{1}{T}\right)} \quad \text{for all } T \in \mathbb{R}_+,$$

then it is enough to check condition (2) only for positive values of  $k$ .

# Examples: approximations with weights

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Let

$$\sum_{i=1}^n \rho_i = \sum_{j=1}^m \sigma_j = 1;$$

define

$$\alpha_i(T) = T^{\rho_i}, \quad \beta_j(T) = T^{\sigma_j}.$$

These functions define the quasinorms  $|\cdot|_\rho$  and  $|\cdot|_\sigma$  on  $\mathbb{R}^n$  and  $\mathbb{R}^m$  by

$$|\mathbf{y}|_\rho = \max_{i=1,\dots,n} |y_i|^{\frac{1}{\rho_i}} \quad \text{and} \quad |\mathbf{x}|_\sigma = \max_{j=1,\dots,m} |x_j|^{\frac{1}{\sigma_j}}.$$

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We use the notation:  $\widehat{\mathbf{D}}[g; \rho, \sigma]$ ,  $\widehat{\mathbf{W}}[g; \rho, \sigma]$ .

# Nontrivial example

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Let  $\theta_1, \theta_2, \eta \in \mathbb{R}$ . Consider the system

$$\begin{cases} \|\theta_1 q_1 + \theta_2 q_2 - \eta\| \leq g(T), \\ |q_1| \leq \beta_1(T), \\ |q_2| \leq \beta_2(T), \end{cases}$$

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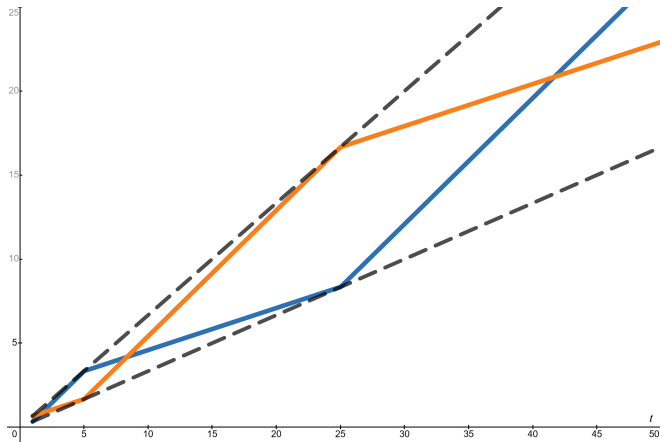
Let  $\theta_1, \theta_2, \eta \in \mathbb{R}$ . Consider the system

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We construct a pair of quasimultiplicative functions  $\beta_1, \beta_2$  such that:

- For an unbounded set of values of  $T$ ,  $\beta_1(T) = T^{\frac{1}{3}}$  and  $\beta_2(T) = T^{\frac{2}{3}}$ ;
- For an unbounded set of values of  $T$ ,  $\beta_1(T) = T^{\frac{2}{3}}$  and  $\beta_2(T) = T^{\frac{1}{3}}$ .

# Nontrivial example



Orange and blue functions are the functions  $\beta_j$  on logarithmic scale (slopes equal  $\frac{3}{4}$  or  $\frac{1}{4}$ ).

# Sections of sets of pairs

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In the inhomogeneous case we can study three types of objects:

- Sets of pairs  $(\Theta, \boldsymbol{\eta})$ , satisfying some restrictions:  $\hat{\mathbf{D}}_{n,m} \left[ g; \underline{\alpha}, \underline{\beta} \right], \hat{\mathbf{W}}_{n,m} \left[ g; \underline{\alpha}, \underline{\beta} \right]$ .
- Sets of matrices  $\Theta$  for a fixed "shift vector"  $\boldsymbol{\eta}$  ("sections with a fixed vector"):  
 $\hat{\mathbf{D}}_{n,m}^{\boldsymbol{\eta}} \left[ g; \underline{\alpha}, \underline{\beta} \right], \hat{\mathbf{W}}_{n,m}^{\boldsymbol{\eta}} \left[ g; \underline{\alpha}, \underline{\beta} \right]$ .
- Sets of vectors  $\boldsymbol{\eta}$  for a fixed matrix  $\Theta$  ("sections with a fixed matrix"):  
 $\hat{\mathbf{D}}_{n,m}^{\Theta} \left[ g; \underline{\alpha}, \underline{\beta} \right], \hat{\mathbf{W}}_{n,m}^{\Theta} \left[ g; \underline{\alpha}, \underline{\beta} \right]$ .

Throughout the talk functions  $\alpha_i, \beta_j$  are quasimultiplicative. Function  $g$  is always a non increasing real valued function.

## **Overview of results**

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# Homogeneous uniform approximation

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We will use the notation:  $f_a(T) = \frac{1}{T^a}$ , in particular,  $f_1(T) = \frac{1}{T}$ .

## Theorem (Dirichlet)

*Any matrix  $\Theta \in \mathbf{M}_{n,m}$  is  $f_1$ -Dirichlet:*

$$\mathbf{D}_{n,m} [f_1] = \mathbf{M}_{n,m}.$$

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Let  $\alpha(T) := \prod_{i=1}^n \alpha_i(T)$  and  $\beta(T) := \prod_{j=1}^m \beta_j(T)$ , and define the function  $g_{\underline{\alpha}, \underline{\beta}}(T) := \alpha^{-1} \left( \frac{1}{\beta(T)} \right)$ .

## Theorem (Simple observation)

Any matrix  $\Theta \in \mathbf{M}_{n,m}$  is  $(g_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta})$ -Dirichlet, and thus is  $(g_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta})$ -approximable.

# Dirichlet improvable matrices have measure zero

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Dirichlet's theorem provides an "for all" bound. It turns out, however, that this bound is also metrically optimal. More precisely:

**Theorem (Davenport, Schmidt, 1970)**

*If  $\varepsilon < 1$ , the set of  $\varepsilon$ -Dirichlet improvable matrices has Lebesgue measure zero:*

$$m(\mathbf{D}_{n,m}[\varepsilon f_1]) = 0.$$

# Transference principle

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From now on, positive decreasing to zero functions  $f$  and  $g$  will always be related by

$$g(T) = \frac{1}{f^{-1}\left(\frac{1}{T}\right)}, \quad f(T) = \frac{1}{g^{-1}\left(\frac{1}{T}\right)}.$$

## Remark

The faster  $f$  decreases, the slower  $g$  decreases.

It is due to Jarnik that:

- The asymptotic approximations of the matrix  $\Theta^\top$  are related to uniform approximations of pairs of the form  $(\Theta, \boldsymbol{\eta})$ .
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# New results: inhomogeneous uniform approximation

## Theorem (Jarnik+)

Suppose the transposed matrix  $\Theta^\top$  is not  $f$ -approximable, that is,  $\Theta^\top \notin \mathbf{W}_{m,n}[f]$ . Then, any pair  $(\Theta, \boldsymbol{\eta})$  is  $g \circ \frac{1}{\kappa^m}$ -Dirichlet, where  $\kappa = \kappa(n, m)$ ; that is,

$$\widehat{\mathbf{D}}_{n,m}^\Theta \left[ g \circ \frac{1}{\kappa^m} \right] = \mathbb{R}^n.$$

## Theorem (Moshchevitin, N., 2025)

Suppose  $\Theta^\top$  is an  $f$ -approximable matrix. If  $\varepsilon$  is small enough, then the set

$$\widehat{\mathbf{D}}_{n,m}^\Theta \left[ \varepsilon \cdot g \circ \frac{1}{\varepsilon} \right]$$

has Lebesgue measure zero. *Works also for intersections with almost every subspace.*

# New results: inhomogeneous uniform approximation

## Theorem (Jarnik+, 2025)

Suppose  $f$  is such a continuous, strictly decreasing function that  $\Theta^T \notin \mathbf{W}_{m,n}[f; \underline{\beta}, \underline{\alpha}]$ . Let  $C = C_d := d \cdot d!$ , where  $d = n + m$ . Then,

$$\hat{\mathbf{D}}_{n,m}^{\Theta} \left[ g; C\underline{\alpha}, C\underline{\beta} \right] = \mathbb{R}^n.$$

## Theorem (N., 2025)

Suppose  $f$  is such that  $\Theta^T \in \mathbf{W}_{m,n}[f; \underline{\beta}, \underline{\alpha}]$ . If  $\varepsilon$  is small enough, then the set

$$\hat{\mathbf{D}}_{n,m}^{\Theta} \left[ \varepsilon \cdot g \circ \frac{1}{\varepsilon}; \underline{\alpha}, \underline{\beta} \right]$$

has Lebesgue measure zero. *Works also for intersections with almost every subspace.*

# New results: inhomogeneous uniform approximation

For the next statement we introduce a natural generalization of singularity.

## Definition

$$\widehat{\mathbf{Sing}}_{n,m}[g] = \bigcap_{\varepsilon > 0} \widehat{\mathbf{D}}[\varepsilon g].$$

The "Twisted Davenport-Schmidt's theorem" shows: if  $\Theta^\top$  is  $f$ -approximable, then the set  $\widehat{\mathbf{Sing}}_{n,m}^\Theta[g]$  has measure zero. The statement below strengthens this statement.

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## Theorem (Moshchevitin, N., 2025)

*Suppose  $f$  satisfies an additional technical condition, and  $\Theta^\top$  is an  $f$ -approximable matrix. Then the set*

$$\left( \widehat{\mathbf{Sing}}_{n,m}^\Theta[g] \right)^c$$

*is HAW; in particular, is winning in Schmidt's game.*

## Dirichlet's theorem for pairs

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# Inhomogeneous Dirichlet's theorem

Theorem (Kleinbock and Wadleigh, 2019)

*The set  $\hat{\mathbf{D}}_{n,m}[g]$  has zero measure if the following series diverges, and full measure if it converges:*

$$\sum_{j=1}^{\infty} \frac{1}{j^2 g(j)}$$

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## Theorem (N., 2025)

*The set  $\widehat{\mathbf{D}}_{n,m}[g; \underline{\alpha}, \underline{\beta}]$  has zero measure if the following series diverges, and full measure if it converges:*

$$\sum_{l=1}^{\infty} \frac{1}{l \cdot \prod_{j=1}^m \beta_j(l) \cdot \prod_{i=1}^n \alpha_i(g(l))}$$

# Khintchine–Groshev's theorem

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## Theorem (Khintchine–Groshev)

*Let  $f(t)$ ,  $t \geq 1$  be a non-increasing real valued function. The set  $\mathbf{W}_{n,m}[f]$  has zero measure if the following series converges, and full measure if it diverges:*

$$\sum_{k=1}^{\infty} f(k)$$

# Khintchine–Groshev’s theorem

## Theorem (Khintchine–Groshev)

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## Theorem (Kleinbock, Wang, 2023)

*Let  $f(T)$ ,  $T \geq 1$  be a non-increasing real valued function. The set  $\mathbf{W}_{m,n}[f; \underline{\beta}, \underline{\alpha}]$  has zero measure if the following series converges, and full measure if it diverges:*

$$\sum_{k=1}^{\infty} k^{-1} \prod_{j=1}^m \beta_j(f(k)) \cdot \prod_{i=1}^n \alpha_i(k)$$

# Proof of equivalence

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- As usual,

$$g(T) = \frac{1}{f^{-1}\left(\frac{1}{T}\right)}, \quad f(T) = \frac{1}{g^{-1}\left(\frac{1}{T}\right)}.$$

- 

$$\int_1^{+\infty} \frac{dz}{z^2 g(z)} \quad \text{and} \quad \int_1^{+\infty} f(t) dt$$

either both converge or both diverge.

## Proof of equivalence: convergent case

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- Suppose  $\sum_{j=1}^{\infty} \frac{1}{j^2 g(j)} < \infty$ . By Khintchine-Groshev's theorem then for almost every  $\Theta$ ,

$$\Theta^{\top} \notin \mathbf{W}_{n,m} [f];$$

and thus

$$\hat{\mathbf{D}}_{n,m}^{\Theta} \left[ g \circ \frac{1}{\kappa^m} \right] = \mathbb{R}^n.$$

So almost every pair is  $g \circ \frac{1}{\kappa^m}$ -Dirichlet.

# Proof of equivalence: divergent case

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$$\Theta^{\top} \in \mathbf{W}_{n,m} [f];$$

and so

$$\hat{\mathbf{D}}_{n,m}^{\Theta} \left[ \varepsilon \cdot g \circ \frac{1}{\varepsilon} \right]$$

has zero measure. Therefore almost no pairs are  $\varepsilon \cdot g \circ \frac{1}{\varepsilon}$ -Dirichlet.

# What happens when measure is zero

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We will complement our metric result by a statement of the following type: *when the function  $g$  decreases so fast that the measure of the sets  $\hat{\mathbf{D}}_{n,m} [g; \underline{\alpha}, \underline{\beta}]$  is zero, these sets are still nonempty and big in certain sense.*

**Theorem (Hong, N. - work in progress)**

*For any decreasing function  $g$ , the set*

$$\hat{\mathbf{D}}_{n,m} [g; \underline{\alpha}, \underline{\beta}]$$

*is uncountable and dense in  $\mathbf{M}_{n,m} \times \mathbb{R}^n$ .*

## **On a criterion of bad approximability**

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# Badly approximable matrices

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Badly approximable matrices are those for which the asymptotic Dirichlet's theorem cannot be improved more than by a multiplicative constant:

$$\mathbf{BA}_{n,m} = \{\Theta \in \mathbf{M}_{n,m} : \|\Theta \mathbf{q}\|_{\mathbb{Z}^n}^n > \frac{c}{|\mathbf{q}|^m} \text{ for some } c > 0 \text{ and any } \mathbf{q} \in \mathbb{Z}^m\}.$$

Despite being of measure zero, badly approximable matrices form a big set in the following sense:

## Theorem (Schmidt + )

*The set  $\mathbf{BA}_{n,m}$  has Lebesgue measure zero. This set is winning in certain games  $((\alpha, \beta)$ -game, hyperplane absolute game...). In particular, this set is dense and has full Hausdorff dimension.*

# Badly approximable matrices: weight functions

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Basing on Dirichlet's theorem for weight functions, we introduce the following natural definition:

## Definition

Matrix  $\Theta \in \mathbf{M}_{n,m}$  is called  $(\underline{\alpha}, \underline{\beta})$ -badly approximable,  $\mathbf{BA}_{n,m}[\underline{\alpha}, \underline{\beta}]$ , if it is not  $(b \cdot g_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta})$ -approximable for some constant  $b > 0$ .

If  $\alpha_i, \beta_j$  are quasimultiplicative, then by generalized Khintchine-Groshev's theorem the set  $\mathbf{BA}_{n,m}[\underline{\alpha}, \underline{\beta}]$  is a Lebesgue nullset.

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Basing on Dirichlet's theorem for weight functions, we introduce the following natural definition:

## Definition

Matrix  $\Theta \in \mathbf{M}_{n,m}$  is called  $(\underline{\alpha}, \underline{\beta})$ -badly approximable,  $\mathbf{BA}_{n,m}[\underline{\alpha}, \underline{\beta}]$ , if it is not  $(b \cdot g_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta})$ -approximable for some constant  $b > 0$ .

If  $\alpha_i, \beta_j$  are quasimultiplicative, then by generalized Khintchine-Groshev's theorem the set  $\mathbf{BA}_{n,m}[\underline{\alpha}, \underline{\beta}]$  is a Lebesgue nullset.

## Proposition

Let  $\underline{\alpha}$  and  $\underline{\beta}$  be tuples of quasimultiplicative functions. Suppose  $\Theta \in \mathbf{M}_{n,m}$  is  $(\underline{\alpha}, \underline{\beta})$ -badly approximable. Then,  $\Theta^\top$  is  $(\underline{\beta}, \underline{\alpha})$ -badly approximable.

# Criterion of bad approximability

## Corollary

Let  $\underline{\alpha}$  and  $\underline{\beta}$  be tuples of quasimultiplicative functions.

- If a matrix  $\Theta \in \mathbf{M}_{n,m}$  is  $(\underline{\alpha}, \underline{\beta})$ -badly approximable, then there exists  $k > 0$  such that any pair  $(\Theta, \boldsymbol{\eta})$  is  $(kg_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta})$ -Dirichlet:

$$\hat{\mathbf{D}}^{\Theta}[kg_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta}] = \mathbb{R}^n.$$

- If a matrix  $\Theta \in \mathbf{M}_{n,m}$  is not  $(\underline{\alpha}, \underline{\beta})$ -badly approximable, then for any  $k > 0$  almost any pair  $(\Theta, \boldsymbol{\eta})$  is not  $(kg_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta})$ -Dirichlet:

$$m\left(\hat{\mathbf{D}}^{\Theta}[kg_{\underline{\alpha}, \underline{\beta}}, \underline{\alpha}, \underline{\beta}]\right) = 0.$$

# Criterion of bad approximability

## Corollary

- If a matrix  $\Theta \in \mathbf{M}_{n,m}$  is badly approximable, then there exists  $k > 0$  such that any pair  $(\Theta, \boldsymbol{\eta})$  is  $kf_1$ -Dirichlet:

$$\hat{\mathbf{D}}^\Theta[kf_1] = \mathbb{R}^n.$$

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$$m\left(\hat{\mathbf{D}}^\Theta[k f_1]\right) = 0.$$

In addition, the complement of  $\hat{\mathbf{D}}^\Theta[k f_1]$  is a HAW set (winning in certain games; in particular Schmidt's game).

# Very good approximability

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Let  $\rho$  and  $\sigma$  be weights.

## Definition

Matrix  $\Theta \in \mathbb{M}_{n,m}$  is called  $(\rho, \sigma)$ -*very well approximable* if it is  $(f_{1+\varepsilon}; \rho, \sigma)$ -approximable for some  $\varepsilon > 0$ .

# Very good approximability

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## Definition

Matrix  $\Theta \in \mathbf{M}_{n,m}$  is called  $(\rho, \sigma)$ -very well approximable if it is  $(f_{1+\varepsilon}; \rho, \sigma)$ -approximable for some  $\varepsilon > 0$ .

## Theorem

- If a matrix  $\Theta \in \mathbf{M}_{n,m}$  is  $(\rho, \sigma)$ -very well approximable, then there exists  $\varepsilon > 0$  such that almost no pair  $(\Theta, \boldsymbol{\eta})$  is  $[f_{1-\varepsilon}; \rho, \sigma]$ -Dirichlet:*

$$m\left(\hat{\mathbf{D}}^{\Theta}[f_{1-\varepsilon}; \rho, \sigma]\right) = 0.$$

- If  $\Theta$  is not  $(\rho, \sigma)$ -very well approximable, then for any  $\varepsilon > 0$  and any  $\boldsymbol{\eta} \in \mathbb{R}^n$  the pair  $(\Theta, \boldsymbol{\eta})$  is  $[f_{1-\varepsilon}; \rho, \sigma]$ -Dirichlet:*

$$\hat{\mathbf{D}}^{\Theta}[f_{1-\varepsilon}; \rho, \sigma] = \mathbb{R}^n.$$

## **Proof of "twisted Davenport-Schmidt's" theorem**

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# Setup

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By assumption,  $\Theta^T \in \mathbf{W}_{m,n}[f; \underline{\beta}, \underline{\alpha}]$ , which means that there exists a sequence  $\mathbf{y}_\nu$  of vectors in  $\mathbb{Z}^n$  for which

$$\|\Theta^T \mathbf{y}_\nu\|_{\underline{\beta}} \leq f(|\mathbf{y}_\nu|_{\underline{\alpha}}); \quad (3)$$

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we will use the notation  $Y_\nu = |\mathbf{y}_\nu|_{\underline{\alpha}}$ .

By quasimultiplicativity, for any  $\delta > 0$  there exists  $\varepsilon = \varepsilon(\delta) > 0$  such that

$$\alpha_i(T) \alpha_i\left(\frac{\varepsilon}{T}\right) \leq \delta \quad \text{and} \quad \beta_j(\varepsilon T) \beta_j\left(\frac{1}{T}\right) \leq \delta \quad \text{for any } i, j \text{ and any } T \in \mathbb{R}_+. \quad (4)$$

Let  $\tilde{g} = \tilde{g}_{\varepsilon(\delta)} = \varepsilon \cdot g \circ \frac{1}{\varepsilon}$ .

# Outline of the proof

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The proof of "twisted Davenport-Schmidt's" theorem consists of two parts:

**Part 1.** Use transference principle to show that the set

$$\widehat{\mathbf{D}}_{n,m}^{\Theta}[\tilde{g}; \underline{\alpha}, \underline{\beta}]$$

is contained in a certain liminf set of a sequence  $\{\Omega_{\nu}\}$  of sets which is explicitly described via  $\{\mathbf{y}_{\nu}\}$ .

**Part 2.** Show: if  $\delta$  is small enough, the sets  $\Omega_{\nu}$  are "almost independent" and deduce an upper bound of  $m \left( \liminf_{\nu \rightarrow \infty} \Omega_{\nu} \right)$ .

# Part 1: transference

---

We define

$$\Omega_\nu = \Omega_\nu((m+n)\delta) := \{\boldsymbol{\eta} \in [0,1)^n : \|\boldsymbol{\eta} \cdot \mathbf{y}_\nu\| \leq (m+n)\delta\}.$$

## Lemma

*Let  $\delta > 0$  be a parameter, and  $\mathbf{y}_\nu, \tilde{g}$  and  $\Omega_\nu$  be chosen as above. Then,*

$$\hat{\mathbf{D}}_{n,m}^\Theta[\tilde{g}; \underline{\alpha}, \underline{\beta}] \cap [0,1)^n \subseteq \bigcup_{N=1}^{\infty} \bigcap_{\nu=N}^{\infty} \Omega_\nu.$$

# Part 1: transference

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Suppose

$$\boldsymbol{\eta} \in \widehat{\mathbf{D}}_{n,m}^{\Theta}[\tilde{g}; \underline{\alpha}, \underline{\beta}] \cap [0, 1)^n,$$

and let  $T_{\nu} = \frac{\varepsilon}{f(Y_{\nu})}$ ; for  $\nu$  large enough there exists  $\mathbf{q}_{\nu}$  such that

$$\begin{cases} |q_{\nu,j}| \leq \beta_j(T_{\nu}), & j = 1, \dots, m; \\ \|\Theta_i \mathbf{q}_{\nu} - \eta_i\| \leq \alpha_i(\tilde{g}(T_{\nu})), & i = 1, \dots, n. \end{cases}$$

# Part 1: transference

---

Classical transference observation:

$$\boldsymbol{\eta} \cdot \mathbf{y}_\nu = \sum_{j=1}^m q_j \Theta_j^T(\mathbf{y}_\nu) - \sum_{i=1}^n (\Theta_i(\mathbf{q}) - \eta_i) y_{\nu,i}.$$

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Bound both terms on the right hand side:

$$\|q_j \Theta_j^T(\mathbf{y}_\nu)\| \leq |q_j| \cdot \|\Theta_j^T(\mathbf{y}_\nu)\| \leq \beta_j(T_\nu) \beta_j(f(Y_\nu)) = \beta_j\left(\frac{\varepsilon}{f(Y_\nu)}\right) \beta_j(f(Y_\nu))$$

and

$$\|(\Theta_i(\mathbf{q}) - \eta_i) y_{\nu,i}\| \leq |y_{\nu,i}| \cdot \|(\Theta_i(\mathbf{q}) - \eta_i)\| \leq \alpha_i(Y_\nu) \alpha_i(\tilde{g}(T_\nu)) = \alpha_i(Y_\nu) \alpha_i\left(\frac{\varepsilon}{Y_\nu}\right).$$

# Part 1: transference

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Combine these estimates: by the choice of  $\varepsilon$  we get that

$$\|\boldsymbol{\eta} \cdot \mathbf{y}_\nu\| \leq \sum_{j=1}^m \|q_j \Theta_j^T(\mathbf{y}_\nu)\| + \sum_{i=1}^n \|(\Theta_i(\mathbf{q}) - \eta_i) y_{\nu,i}\| \leq (m+n)\delta.$$

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The condition above is  $\boldsymbol{\eta} \in \Omega_\nu$  and it should hold for all  $\nu$  **large enough**, so

$$\boldsymbol{\eta} \in \bigcup_{N=1}^{\infty} \bigcap_{\nu=N}^{\infty} \Omega_\nu.$$

## Part 2: measure estimate

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### Lemma

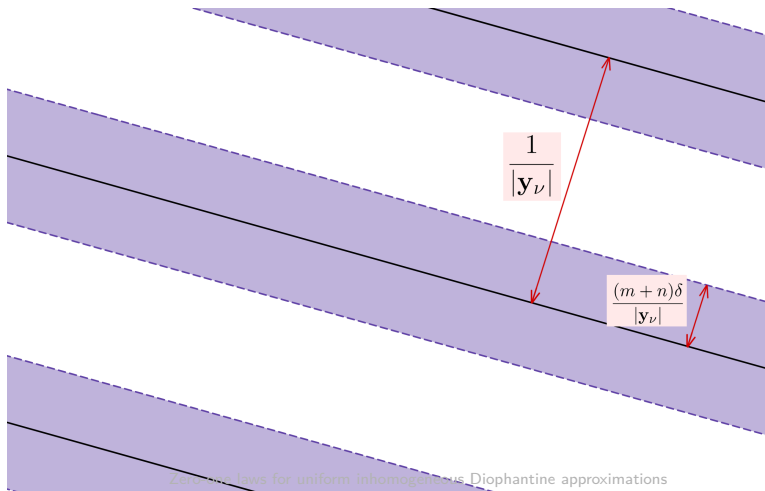
Let  $0 < \delta < \frac{1}{2(n+m)}$ . Then,

$$m\left(\bigcup_{N=1}^{\infty} \bigcap_{\nu=N}^{\infty} \Omega_{\nu}\right) = 0.$$

First, we want to switch from  $\{\mathbf{y}_{\nu}\}$  to a subsequence with an additional condition  $\lim_{\nu \rightarrow \infty} \frac{|\mathbf{y}_{\nu+1}|}{|\mathbf{y}_{\nu}|} = \infty$ . It can only increase the liminf set we are working with.

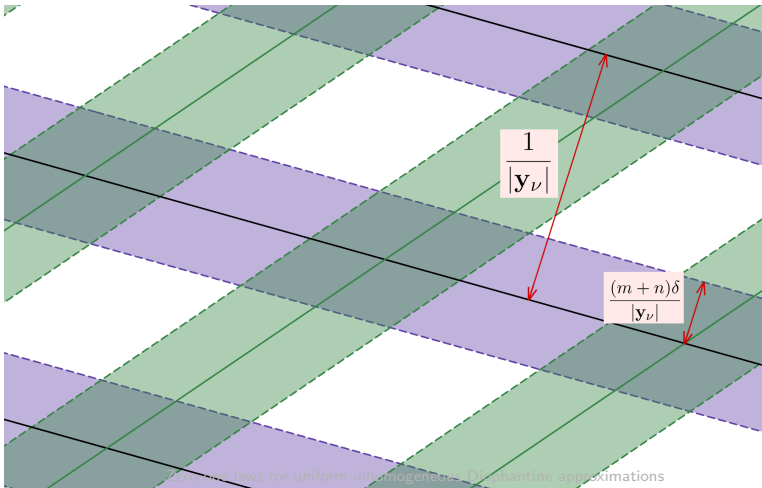
## Part 2: measure estimate

The sets  $\Omega_\nu = \{\boldsymbol{\eta} \in [0, 1)^n : ||\boldsymbol{\eta} \cdot \mathbf{y}_\nu|| \leq (m+n)\delta\}$  consist of "stripes" —  $\frac{(m+n)\delta}{|\mathbf{y}_\nu|}$ -neighborhoods of the hyperplanes  $\boldsymbol{\eta} \cdot \mathbf{y}_\nu = p$  where  $p \in \mathbb{Z}$ :



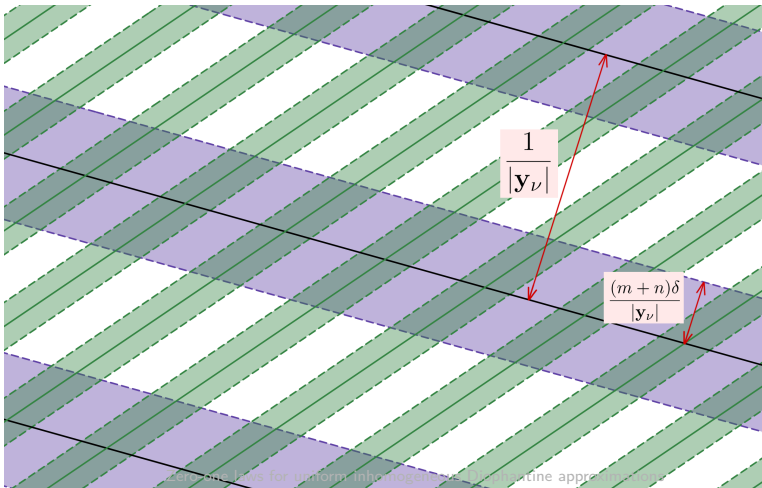
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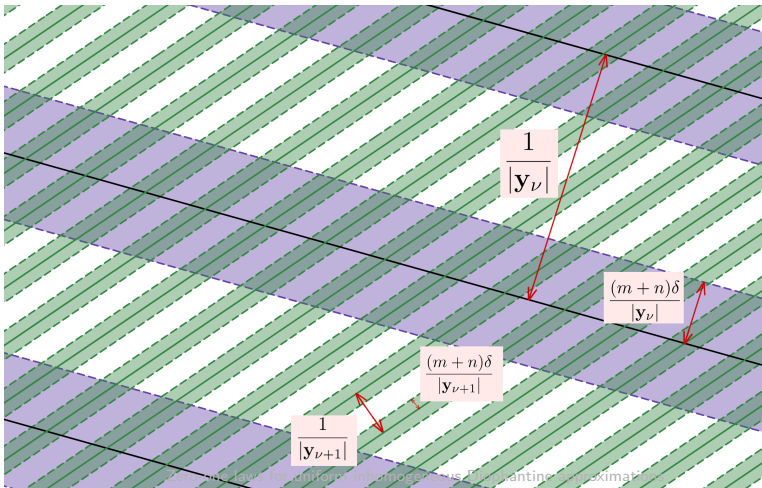
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## Part 2: measure estimate

---

**Claim.** For  $\nu \neq \mu$  large enough,

$$m\left(\Omega_\nu^c \cap \Omega_\mu^c\right) \leq (1 + \lambda) m\left(\Omega_\nu^c\right) m\left(\Omega_\mu^c\right).$$

## Part 2: measure estimate

**Claim.** For  $\nu \neq \mu$  large enough,

$$m\left(\Omega_\nu^c \cap \Omega_\mu^c\right) \leq (1 + \lambda) m\left(\Omega_\nu^c\right) m\left(\Omega_\mu^c\right).$$

This allows us to apply the following classical statement to  $\limsup_{\nu \rightarrow \infty} \Omega_\nu^c$ :

### Lemma

Let  $X$  be a measure space with measure  $\mu$ , and let  $\{E_k\}_{k=1}^\infty$  be a countable collection of measurable sets. Let  $E = \bigcap_{N=1}^\infty \bigcup_{k=N}^\infty E_k$ . Suppose  $\sum_{k=1}^\infty \mu(E_k) = \infty$ . Then,

$$\mu(E) \geq \limsup_{N \rightarrow \infty} \frac{\left(\sum_{k=1}^N \mu(E_k)\right)^2}{\sum_{k,s=1}^N \mu(E_k \cap E_s)}.$$

## Exceptional subspaces

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# $\Xi$ -Exceptional subspaces

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Let  $\{\mathbf{y}_\nu\}$  be the sequence used in the proof of "inhomogeneous Davenport-Schmidt's" theorem.

## Definition

We call a point  $\mathbf{v} \in \mathbb{R}^n$  on the unit sphere (direction)  $\Xi$ -*exceptional* (for the matrix  $\Theta^\top$  and function  $f$ ) if for any  $\delta > 0$  there exist infinitely many  $\nu \in \mathbb{N}$  for which  $\widehat{\mathbf{y}_\nu, \mathbf{v}} < \frac{\delta}{|\mathbf{y}_\nu|^\Xi}$ .

## Definition

Let  $\mathcal{A}$  be an affine subspace of  $\mathbb{R}^n$  and  $\mathcal{L}$  — the corresponding linear subspace. We call  $\mathcal{A}$   $\Xi$ -*exceptional* if  $\mathcal{L}^\perp$  contains  $\Xi$ -exceptional points.

# $\Xi$ -Exceptional subspaces

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Informally, non-exceptional subspaces are those where the projections of  $\mathbf{y}_\nu$  grow fast enough.

## Lemma

Let  $\Xi > 0$  and  $X_E$  be the set of  $k$ -dimensional  $\Xi$ -exceptional linear subspaces of  $\mathbb{R}^n$ . Then,  $\dim_H X_E = k(n - k - 1)$ , if not empty.

In particular, almost no affine subspaces are  $\Xi$ -exceptional.

The "inhomogeneous Davenport-Schmidt theorem" works for  $\boldsymbol{\eta}$  restricted on any non-exceptional subspace of  $\mathbb{R}^n$ .

**Thank you for your attention!**

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