

Optimal order digital nets and sequences

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Joint work with Kosuke Suzuki and Takehito Yoshiki

RICAM Discrepancy Workshop, November 2018

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- BUT about how the idea from discrepancy problem can be used to obtain some results in numerical integration.

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- BUT about how the idea from discrepancy problem can be used to obtain some results in numerical integration.
- The following is a list of our papers which this talk is based on.
 - ▶ TG, K. Suzuki, T. Yoshiki: Optimal order quadrature error bounds for infinite-dimensional higher-order digital sequences, *Found. Comput. Math.* 18 (2018) 433–458.
 - ▶ TG, K. Suzuki, T. Yoshiki: Optimal order quasi-Monte Carlo integration in weighted Sobolev spaces of arbitrary smoothness, *IMA J. Numer. Anal.* 37 (2017), 505–518.
 - ▶ TG, K. Suzuki, T. Yoshiki: An explicit construction of optimal order quasi-Monte Carlo rules for smooth integrands, *SIAM J. Numer. Anal.* 54 (2016), 2664–2683.

L_p -discrepancy

Definition

For an N -element point set $P \subset [0, 1]^s$, the L_p -discrepancy is defined by

$$(L_p(P))^p := \int_{[0,1]^s} \left| \frac{1}{N} \sum_{\mathbf{x} \in P} \mathbf{1}_{\mathbf{x} \in [0, \mathbf{y}]} - \lambda([0, \mathbf{y}]) \right|^p d\mathbf{y},$$

where λ denotes the Lebesgue measure in \mathbb{R}^s .

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where λ denotes the Lebesgue measure in \mathbb{R}^s .

- In case of $p = 2$,

$$(L_2(P))^2 = \frac{1}{3^s} - \frac{2}{N} \sum_{\mathbf{x} \in P} \prod_{j=1}^s \frac{1 - x_j^2}{2} + \frac{1}{N^2} \sum_{\mathbf{x}, \mathbf{y} \in P} \prod_{j=1}^s \min(1 - x_j, 1 - y_j).$$

Quasi-Monte Carlo (QMC) integration

Problem

Approximate/Estimate

$$I(f) := \int_{[0,1]^s} f(\mathbf{x}) \, d\mathbf{x},$$

where $s \in \mathbb{N}$ and $f: [0, 1]^s \rightarrow \mathbb{R}$ is integrable.

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- Choose an N -element point set $P \subset [0, 1]^s$.
- Approximate $I(f)$ by

$$Q_P(f) := \frac{1}{N} \sum_{\mathbf{x} \in P} f(\mathbf{x}).$$

- For an infinite sequence of points $\mathcal{S} = \{\mathbf{x}_n \mid n \geq 0\}$, the first N elements of \mathcal{S} are used as P .

Worst-case error

Definition

For a function space V with norm $\|\cdot\|_V$,

$$e^{\text{wor}}(V, P) := \sup_{\|f\|_V \leq 1} |Q_P(f) - I(f)|.$$

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For a function space V with norm $\|\cdot\|_V$,

$$e^{\text{wor}}(V, P) := \sup_{\|f\|_V \leq 1} |Q_P(f) - I(f)|.$$

- We want to construct a good P depending on V .
- When V is an RKHS with kernel K ,

$$\begin{aligned} (e^{\text{wor}}(V, P))^2 &= \int_{[0,1]^s} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{x} \, d\mathbf{y} \\ &\quad - \frac{2}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} + \frac{1}{N^2} \sum_{\mathbf{x}, \mathbf{y} \in P} K(\mathbf{x}, \mathbf{y}). \end{aligned}$$

L_2 -discrepancy (again)

- Let us consider an RKHS with

$$K_{1,s}^*(\mathbf{x}, \mathbf{y}) = \prod_{j=1}^s \min(1 - x_j, 1 - y_j).$$

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The L_2 -discrepancy of P is same as the worst-case error in this RKHS.

- This RKHS is

$$V = H_{1,s}^* = \bigotimes_{j=1}^s H_1^*$$

where

$$H_1^* = \left\{ f: [0, 1] \rightarrow \mathbb{R} \mid f(1) = 0, f^{(1)} \in L_2 \right\}.$$

Optimal order QMC for L_2 -discrepancy

- QMC point sets, which achieve the optimal order worst-case error in this RKHS, also achieve the optimal order L_2 -discrepancy:

$$L_2(P) \asymp \frac{(\log N)^{(s-1)/2}}{N}.$$

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- For $s = 1$ and $s = 2$, there are many explicit constructions of such point sets.
- But for $s \geq 3$, we only know the results from
 - ① Chen and Skriganov (2002), Skriganov (2006)
 - ② Dick and Pillichshammer (2014), Dick, Hinrichs, Markhasin and Pillichshammer (2017): **higher order digital nets/sequences**
 - ③ Levin (2018): Halton sequences

Sobolev spaces of our interest

- In this work we consider functions of higher smoothness:

$$V = H_{\alpha,s} = \bigotimes_{j=1}^s H_{\alpha},$$

where $\alpha \geq 2$ and

$$H_{\alpha} = \left\{ f : [0, 1] \rightarrow \mathbb{R} \mid f^{(r)} : \text{abs. conti. for } 0 \leq r \leq \alpha - 1, f^{(\alpha)} \in L_2 \right\}.$$

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- $H_{\alpha,s}$ coincides with an RKHS with kernel

$$K_{\alpha,s}(\mathbf{x}, \mathbf{y}) = \prod_{j=1}^s \left[\sum_{r=0}^{\alpha} \frac{B_r(x_j)B_r(y_j)}{(r!)^2} + (-1)^{\alpha+1} \frac{B_{2\alpha}(|x_j - y_j|)}{(2\alpha)!} \right],$$

where B_r denotes the Bernoulli poly. of degree r (Wahba, 1990).

Higher order convergence

Known results (Dick, 2008; Baldeaux and Dick, 2009)

Higher order digital nets/sequences achieve

$$e^{\text{wor}}(H_{\alpha,s}, P) \ll \frac{(\log N)^{c(\alpha,s)}}{N^\alpha}.$$

- $c(\alpha, s) = \alpha s$ is obtained for order α digital nets.
- The best possible exponent of $\log N$ term is $(s - 1)/2$ (for linear quadrature algorithm).

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Aim of this talk

Prove that higher order digital nets/sequences achieve the *optimal* order of convergence

$$e^{\text{wor}}(H_{\alpha,s}, P) \ll \frac{(\log N)^{(s-1)/2}}{N^\alpha}.$$

Main result

Theorem (G., Suzuki and Yoshiki, 2018)

For $\alpha \geq 2$, order $2\alpha + 1$ digital nets/sequences in prime base b achieve

$$e^{\text{wor}}(H_{\alpha,s}, P) \asymp \frac{(\log N)^{(s-1)/2}}{N^\alpha}$$

for $N = b^m$ with $m \in \mathbb{N}$.

In the rest of this talk

- I focus on
 - 1 nets, and
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- I focus on
 - 1 nets, and
 - 2 an upper bound. (**Remark:** A lower bound follows from the “bumping function” argument by Bakhvalov (1959).)
- I want to
 - 1 introduce higher order digital nets, and
 - 2 show a sketch of our proof for the main result by highlighting an analogy to the proof by Dick and Pillichshammer (2014), who proved the optimal order L_2 -discrepancy bound for order 3 digital nets.

Digital nets

Definition (Niederreiter, 1992)

- For prime b , \mathbb{F}_b denotes the b -element field.

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- For $0 \leq h < b^m$, write $h = \eta_0 + \eta_1 b + \dots + \eta_{m-1} b^{m-1}$.
- For $1 \leq j \leq s$, let

$$x_{h,j} = \frac{\xi_{1,h,j}}{b} + \dots + \frac{\xi_{n,h,j}}{b^n} \in [0, 1],$$

where

$$(\xi_{1,h,j}, \dots, \xi_{n,h,j})^\top = C_j \cdot (\eta_0, \dots, \eta_{m-1})^\top \in \mathbb{F}_b^n.$$

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$$(\xi_{1,h,j}, \dots, \xi_{n,h,j})^\top = C_j \cdot (\eta_0, \dots, \eta_{m-1})^\top \in \mathbb{F}_b^n.$$

- Set $\mathbf{x}_h = (x_{h,1}, \dots, x_{h,s}) \in [0, 1]^s$.

We call $P = \{\mathbf{x}_h : 0 \leq h < b^m\}$ a *digital net* over \mathbb{F}_b with generating matrices C_1, \dots, C_s .

Dual net

Definition

The dual net of P with C_1, \dots, C_s is defined by

$$P^\perp = \left\{ (k_1, \dots, k_s) \in \mathbb{N}_0^s : C_1^\top \vec{k}_1 \oplus \dots \oplus C_s^\top \vec{k}_s = \mathbf{0} \in \mathbb{F}_b^m \right\},$$

where we write

$$\vec{k} = (\kappa_0, \dots, \kappa_{n-1}) \in \mathbb{F}_b^n$$

for $k = \kappa_0 + \kappa_1 b + \dots$.

NRT metric and (t, m, s) -nets

- For $k = \kappa_1 b^{a_1-1} + \dots + \kappa_v b^{a_v-1}$ with $\kappa_i \in \{1, \dots, b-1\}$ and $a_1 > a_2 > \dots$, the NRT metric is

$$\mu_1(k) = a_1 \quad \text{and} \quad \mu_1(k_1, \dots, k_s) = \sum_{j=1}^s \mu_1(k_j).$$

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Definition

We call P a digital (t, m, s) -net if

$$\mu_1(P^\perp) := \min_{\mathbf{k} \in P^\perp \setminus \{\mathbf{0}\}} \mu_1(\mathbf{k}) \geq m - t + 1.$$

(Constructions: Sobol', Faure, Niederreiter, Tezuka, Niederreiter-Xing, ...)

Dick metric and higher order nets

- Let $\alpha \in \mathbb{N}$.
- For $k = \kappa_1 b^{a_1-1} + \dots + \kappa_v b^{a_v-1}$ with $\kappa_j \in \{1, \dots, b-1\}$ and $a_1 > a_2 > \dots$, the Dick metric is

$$\mu_\alpha(k) = a_1 + \dots + a_{\min(v, \alpha)} \quad \text{and} \quad \mu_\alpha(k_1, \dots, k_s) = \sum_{j=1}^s \mu_\alpha(k_j).$$

$\alpha = 1$: the NRT metric.

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$\alpha = 1$: the NRT metric.

Definition

We call P an order α digital (t, m, s) -net if

$$\mu_\alpha(P^\perp) := \min_{\mathbf{k} \in P^\perp \setminus \{\mathbf{0}\}} \mu_\alpha(\mathbf{k}) \geq \alpha m - t + 1.$$

Explicit construction

- Define $\mathcal{D}_\alpha : [0, 1]^\alpha \rightarrow [0, 1]$ by

$$\begin{cases} x_1 = (0.\xi_{1,1}\xi_{2,1}\xi_{3,1}\dots)_b \\ x_2 = (0.\xi_{1,2}\xi_{2,2}\xi_{3,2}\dots)_b \\ \vdots \\ x_\alpha = (0.\xi_{1,\alpha}\xi_{2,\alpha}\xi_{3,\alpha}\dots)_b \end{cases} \mapsto (0.\underbrace{\xi_{1,1}\xi_{1,2}\dots\xi_{1,\alpha}}_\alpha \underbrace{\xi_{2,1}\xi_{2,2}\dots\xi_{2,\alpha}}_\alpha \dots)_b.$$

- For $\mathbf{x} \in [0, 1]^{\alpha s}$, we write

$$\mathcal{D}_\alpha(\mathbf{x}) = (\mathcal{D}_\alpha(x_1, \dots, x_\alpha), \dots, \mathcal{D}_\alpha(x_{\alpha(s-1)+1}, \dots, x_{\alpha s})) \in [0, 1]^s.$$

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- For $\mathbf{x} \in [0, 1]^{\alpha s}$, we write

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Theorem (Dick, 2007)

Let P be a digital $(t, m, \alpha s)$ -net. Then $\mathcal{D}_\alpha(P)$ is an order α digital (t_α, m, s) -net with $t_\alpha \leq \alpha \min \{m, t + \lfloor s(\alpha - 1)/2 \rfloor\}$.

Another look at construction

- Let P be a digital $(t, m, \alpha s)$ -net with $C_1, \dots, C_{\alpha s} \in \mathbb{F}_b^{m \times m}$. We write

$$C_1 = \begin{pmatrix} \mathbf{c}_{1,1} \\ \vdots \\ \mathbf{c}_{m,1} \end{pmatrix}, \dots, C_\alpha = \begin{pmatrix} \mathbf{c}_{1,\alpha} \\ \vdots \\ \mathbf{c}_{m,\alpha} \end{pmatrix}, C_{\alpha+1} = \begin{pmatrix} \mathbf{c}_{1,\alpha+1} \\ \vdots \\ \mathbf{c}_{m,\alpha+1} \end{pmatrix} \dots$$

Another look at construction

- Let P be a digital (t, m, α) -net with $C_1, \dots, C_{\alpha s} \in \mathbb{F}_b^{m \times m}$. We write

$$C_1 = \begin{pmatrix} \mathbf{c}_{1,1} \\ \vdots \\ \mathbf{c}_{m,1} \end{pmatrix}, \dots, C_{\alpha} = \begin{pmatrix} \mathbf{c}_{1,\alpha} \\ \vdots \\ \mathbf{c}_{m,\alpha} \end{pmatrix}, C_{\alpha+1} = \begin{pmatrix} \mathbf{c}_{1,\alpha+1} \\ \vdots \\ \mathbf{c}_{m,\alpha+1} \end{pmatrix} \dots$$

- $\mathcal{D}_{\alpha}(P)$ is a digital net with $D_1, \dots, D_s \in \mathbb{F}_b^{\alpha m \times m}$ where

$$D_1 = \begin{pmatrix} \mathbf{c}_{1,1} \\ \vdots \\ \mathbf{c}_{1,\alpha} \\ \vdots \\ \mathbf{c}_{m,1} \\ \vdots \\ \mathbf{c}_{m,\alpha} \end{pmatrix}, D_2 = \begin{pmatrix} \mathbf{c}_{1,\alpha+1} \\ \vdots \\ \mathbf{c}_{1,2\alpha} \\ \vdots \\ \mathbf{c}_{m,\alpha+1} \\ \vdots \\ \mathbf{c}_{m,2\alpha} \end{pmatrix}, \dots$$

Toward the proof: Walsh functions

Definition

- For $k = \kappa_0 + \kappa_1 b + \dots$, the k -th Walsh function wal_k is defined by

$$\text{wal}_k(x) := \exp \left[\frac{2\pi i}{b} (\kappa_0 \xi_1 + \kappa_1 \xi_2 + \dots) \right],$$

where $x = \xi_1/b + \xi_2/b^2 + \dots$.

- For vectors $\mathbf{k} \in \mathbb{N}_0^s$, we define

$$\text{wal}_{\mathbf{k}}(\mathbf{x}) := \prod_{j=1}^s \text{wal}_{k_j}(x_j).$$

ONB

- The set $\{\text{wal}_{\mathbf{k}} : \mathbf{k} \in \mathbb{N}_0^s\}$ forms an ONB in $L_2([0, 1]^s)$.
- Therefore, in general, we have

$$K(\mathbf{x}, \mathbf{y}) = \sum_{\mathbf{k}, \mathbf{l} \in \mathbb{N}_0^s} \hat{K}(\mathbf{k}, \mathbf{l}) \text{wal}_{\mathbf{k}}(\mathbf{x}) \overline{\text{wal}_{\mathbf{l}}(\mathbf{y})},$$

where $\hat{K}(\mathbf{k}, \mathbf{l})$ is the (\mathbf{k}, \mathbf{l}) -th Walsh coefficient

$$\hat{K}(\mathbf{k}, \mathbf{l}) = \int_{[0,1]^s} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \overline{\text{wal}_{\mathbf{k}}(\mathbf{x}) \text{wal}_{\mathbf{l}}(\mathbf{y})} \, d\mathbf{x} \, d\mathbf{y}.$$

Character property

Lemma

Let P be a digital net with $C_1, \dots, C_s \in \mathbb{F}_b^{n \times m}$. We have

$$\sum_{\mathbf{x} \in P} \text{wal}_{\mathbf{k}}(\mathbf{x}) = \begin{cases} b^m = N & \text{if } \mathbf{k} \in P^\perp, \\ 0 & \text{otherwise.} \end{cases}$$

Worst-case error for digital nets

Lemma

For an RKHS V with kernel K and a digital net P ,

$$(e^{\text{wor}}(V, P))^2 = \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}(\mathbf{k}, \mathbf{l}).$$

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$$(e^{\text{wor}}(V, P))^2 = \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}(\mathbf{k}, \mathbf{l}).$$

Proof: Recall the following explicit formula

$$\begin{aligned} (e^{\text{wor}}(V, P))^2 &= \int_{[0,1]^s} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{x} \, d\mathbf{y} \\ &\quad - \frac{2}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} + \frac{1}{N^2} \sum_{\mathbf{x}, \mathbf{y} \in P} K(\mathbf{x}, \mathbf{y}). \end{aligned}$$

Worst-case error for digital nets (continued)

The first term is

$$\int_{[0,1]^s} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{x} \, d\mathbf{y} = \hat{K}(\mathbf{0}, \mathbf{0}).$$

Worst-case error for digital nets (continued)

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The second term is

$$\begin{aligned} & \frac{2}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} \\ &= \frac{1}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} + \frac{1}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{y}, \mathbf{x}) \, d\mathbf{y} \end{aligned}$$

Worst-case error for digital nets (continued)

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$$\begin{aligned} & \frac{2}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} \\ &= \frac{1}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{x}, \mathbf{y}) \, d\mathbf{y} + \frac{1}{N} \sum_{\mathbf{x} \in P} \int_{[0,1]^s} K(\mathbf{y}, \mathbf{x}) \, d\mathbf{y} \\ &= \frac{1}{N} \sum_{\mathbf{x} \in P} \sum_{\mathbf{k} \in \mathbb{N}_0^s} \hat{K}(\mathbf{k}, \mathbf{0}) \text{wal}_{\mathbf{k}}(\mathbf{x}) + \frac{1}{N} \sum_{\mathbf{x} \in P} \sum_{\mathbf{l} \in \mathbb{N}_0^s} \hat{K}(\mathbf{0}, \mathbf{l}) \overline{\text{wal}_{\mathbf{l}}(\mathbf{x})} \end{aligned}$$

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Worst-case error for digital nets (continued)

Finally, the third term is

$$\frac{1}{N^2} \sum_{\mathbf{x}, \mathbf{y} \in P} K(\mathbf{x}, \mathbf{y}) = \frac{1}{N^2} \sum_{\mathbf{x}, \mathbf{y} \in P} \sum_{\mathbf{k}, \mathbf{l} \in \mathbb{N}_0^s} \hat{K}(\mathbf{k}, \mathbf{l}) \text{wal}_{\mathbf{k}}(\mathbf{x}) \overline{\text{wal}_{\mathbf{l}}(\mathbf{y})}$$

Worst-case error for digital nets (continued)

Finally, the third term is

$$\begin{aligned}\frac{1}{N^2} \sum_{\mathbf{x}, \mathbf{y} \in P} K(\mathbf{x}, \mathbf{y}) &= \frac{1}{N^2} \sum_{\mathbf{x}, \mathbf{y} \in P} \sum_{\mathbf{k}, \mathbf{l} \in \mathbb{N}_0^s} \hat{K}(\mathbf{k}, \mathbf{l}) \text{wal}_{\mathbf{k}}(\mathbf{x}) \overline{\text{wal}_{\mathbf{l}}(\mathbf{y})} \\ &= \sum_{\mathbf{k}, \mathbf{l} \in \mathbb{N}_0^s} \hat{K}(\mathbf{k}, \mathbf{l}) \frac{1}{N} \sum_{\mathbf{x} \in P} \text{wal}_{\mathbf{k}}(\mathbf{x}) \overline{\frac{1}{N} \sum_{\mathbf{y} \in P} \text{wal}_{\mathbf{l}}(\mathbf{y})}\end{aligned}$$

Worst-case error for digital nets (continued)

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Altogether we have

$$\begin{aligned}(e^{\text{wor}}(V, P))^2 &= \hat{K}(\mathbf{0}, \mathbf{0}) - \sum_{\mathbf{k} \in P^\perp} \hat{K}(\mathbf{k}, \mathbf{0}) - \sum_{\mathbf{l} \in P^\perp} \hat{K}(\mathbf{0}, \mathbf{l}) + \sum_{\mathbf{k}, \mathbf{l} \in P^\perp} \hat{K}(\mathbf{k}, \mathbf{l}) \\ &= \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}(\mathbf{k}, \mathbf{l}).\end{aligned}$$

Learning from L_2 -discrepancy for digital nets

- For a digital net P ,

$$(L_2(P))^2 = \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}_{1,s}^*(\mathbf{k}, \mathbf{l}),$$

where

$$K_{1,s}^*(\mathbf{x}, \mathbf{y}) = \prod_{j=1}^s \min(1 - x_j, 1 - y_j).$$

- Constructions of digital nets with optimal order L_2 -discrepancy
 - ▷ Chen and Skriganov (2002)
 - ▷ Dick and Pillichshammer (2014): order 3 digital nets

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Common idea

is to **exploit the decay and the sparsity** of $\hat{K}_{1,s}^*$.

Decay: $\hat{K}_{1,s}^*(\mathbf{k}, \mathbf{l}) \ll b^{-\mu_1(\mathbf{k}) - \mu_1(\mathbf{l})}$

Sparsity: $\hat{K}_{1,s}^*(\mathbf{k}, \mathbf{l}) = 0$ for many (\mathbf{k}, \mathbf{l})

Rough sketch: Optimal order L_2 -discrepancy bounds

P : an order 3 digital (t, m, s) -net

$$(L_2(P))^2 = \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}_{1,s}^*(\mathbf{k}, \mathbf{l})$$

(Dick and Pillichshammer, 2014)

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(Dick and Pillichshammer, 2014)

If only the decay of $\hat{K}_{\alpha,s}$ is considered

$$(e^{\text{wor}}(H_{\alpha,s}, P))^2 = \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}_{\alpha,s}(\mathbf{k}, \mathbf{l}),$$

where

$$K_{\alpha,s}(\mathbf{x}, \mathbf{y}) = \prod_{j=1}^s \left[\sum_{r=0}^{\alpha} \frac{B_r(x_j) B_r(y_j)}{(r!)^2} + (-1)^{\alpha+1} \frac{B_{2\alpha}(|x_j - y_j|)}{(2\alpha)!} \right].$$

Lemma (Baldeaux and Dick, 2009)

For $\mathbf{k}, \mathbf{l} \in \mathbb{N}_0^s$,

$$\hat{K}_{\alpha,s}(\mathbf{k}, \mathbf{l}) \ll b^{-\mu_\alpha(\mathbf{k}) - \mu_\alpha(\mathbf{l})}.$$

Rough sketch: Nearly optimal order error bounds

P : an order α digital (t, m, s) -net

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- The exponent of $\log N$ terms comes from the counting step with respect to μ_α -metric.

Effect of switching the metric

- Counting along μ_α -metric gives

$$\left| \{ \mathbf{k} \in P^\perp \setminus \{ \mathbf{0} \} : \mu_\alpha(\mathbf{k}) = z \} \right| \ll b^{z-\alpha m} m^{\alpha s} \quad \text{for } z \geq \mu_\alpha(P^\perp),$$

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while counting along μ_1 -metric gives

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- We want to
 - switch the metric from μ_α to μ_1 , and
 - exploit the sparsity of $\hat{K}_{\alpha,s}$ to deal with the double sum $\sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{ \mathbf{0} \}}$.

Analogy to L_2 -discrepancy problem

$$(e^{\text{WOR}}(H_{\alpha,s}, P))^2 = \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}_{\alpha,s}(\mathbf{k}, \mathbf{l}).$$

Exploit the decay and the sparsity of $\hat{K}_{\alpha,s}$

Decay: $\hat{K}_{\alpha,s}(\mathbf{k}, \mathbf{l}) \ll b^{-\mu_\alpha(\mathbf{k}) - \mu_\alpha(\mathbf{l})}$ (Baldeaux and Dick, 2009)

Sparsity: $\hat{K}_{\alpha,s}(\mathbf{k}, \mathbf{l}) = 0$ for many (\mathbf{k}, \mathbf{l}) (G., Suzuki and Yoshiki, 2018)

Analogy to L_2 -discrepancy problem $\pm \epsilon$

- In order to switch the metric in the counting step, we need:

Additional tricks

Propagation: P : order $2\alpha + 1$ net $\Rightarrow P$: order 1 net (Dick, 2008)

Interpolation: $\mu_\alpha(\mathbf{k}) \geq A\mu_{2\alpha+1}(\mathbf{k}) + B\mu_1(\mathbf{k})$

where $A = \frac{\alpha - 1}{2\alpha}$ and $B = \frac{\alpha + 1}{2\alpha}$ (G., Suzuki and Yoshiki, 2017).

- The interpolation inequality on μ_α is nothing but Jensen's inequality with respect to α .

Remark: how sparse is $\hat{K}_{\alpha,s}$?

- Let

$$k = \sum_{i=1}^v \kappa_i b^{c_i-1} \quad \text{and} \quad l = \sum_{i=1}^w \lambda_i b^{d_i-1},$$

such that $c_1 > c_2 > \dots$, $d_1 > d_2 > \dots$ and $\kappa_i, \lambda_i \in \{1, \dots, b-1\}$.

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- For $p, q \in \mathbb{N}_0$, we write

$$k^{(p)} = \sum_{i=p+1}^v \kappa_i b^{c_i-1} \quad \text{and} \quad l^{(q)} = \sum_{i=q+1}^w \lambda_i b^{d_i-1}$$

where the empty sum is set to 0.

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- We say “ (k, l) is of type (p, q) ” if $k^{(p)} = l^{(q)}$ and $\kappa_p b^{c_p-1} \neq \lambda_q b^{d_q-1}$.

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Lemma (G., Suzuki and Yoshiki, 2018)

If there exists one index $j \in \{1, \dots, s\}$ such that (k_j, l_j) is of type (p_j, q_j) with $p_j + q_j > 2\alpha$, then $\hat{K}_{\alpha,s}(\mathbf{k}, \mathbf{l}) = 0$.

Rough sketch: Optimal order error bounds

P : an order $2\alpha + 1$ digital (t, m, s) -net

$$(e^{\text{wor}}(H_{\alpha, s}, P))^2 = \sum_{\mathbf{k}, \mathbf{l} \in P^\perp \setminus \{\mathbf{0}\}} \hat{K}_{\alpha, s}(\mathbf{k}, \mathbf{l})$$

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 &\quad \times \left| \{(\mathbf{k}, \mathbf{l}) \in (P^\perp \setminus \{\mathbf{0}\})^2 : \mu_1(\mathbf{k}) + \mu_1(\mathbf{l}) = z, \hat{K}_{\alpha,s}(\mathbf{k}, \mathbf{l}) \neq 0\} \right| \\
 &\ll \frac{m^{s-1}}{b^{2A\mu_{2\alpha+1}(P^\perp) + 2B\mu_1(P^\perp)}} \ll \frac{m^{s-1}}{b^{2\alpha m}} = \frac{(\log_b N)^{s-1}}{N^{2\alpha}}.
 \end{aligned}$$

Some comments

- Again I want to stress similarity of the proofs. In the proof of Dick and Pillichshammer (2014), we see

$$\left| \{(\mathbf{k}, \mathbf{l}) \in (P^\perp \setminus \{\mathbf{0}\})^2 : \mu_1(\mathbf{k}) + \mu_1(\mathbf{l}) = z, \hat{K}_{1,s}^*(\mathbf{k}, \mathbf{l}) \neq 0\} \right|.$$

In our proof, we see

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- On order $2\alpha + 1$ digital *sequences*: The optimal order error bound holds only for a geometric sequence of # of points $N = b^1, b^2, \dots$. This cannot be improved for any extensible QMC rule (Owen, 2016).

Summary

- Order $2\alpha + 1$ digital nets/sequences achieve the optimal order of convergence for numerical integration in $H_{\alpha,s} = \bigotimes_{j=1}^s H_{\alpha}$.
- Explicit construction of higher order digital nets and sequences is possible. We obtain an optimal order quadrature rule which is extensible both in the number of points and the dimension.
- The main idea of the proof comes from the L_2 -discrepancy problem together with some additional tricks (interpolation and propagation).

“Deep” learning from L_2 -discrepancy problem?

- After the work of Dick and Pillichshammer (2014), it has been proven that order 2 (instead of 3) digital sequences achieve the best possible order of L_p -discrepancy for all $1 < p < \infty$ by
 - ▶ Dick, Hinrichs, Markhasin and Pillichshammer (2017) Israel J. Math.
- The proof is based on *Haar functions* instead of Walsh functions.

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- The proof is based on *Haar functions* instead of Walsh functions.
- Can we learn further from their result? Is order $2\alpha + 1$ best possible?? Does Haar analysis help us???
- How about other function spaces (e.g., Besov and Triebel-Lizorkin)? For the case of smoothness less than 2, please refer to
 - ▶ Hinrichs, Markhasin, Oettershagen and Ullrich (2016) Numer. Math.
- We do not have any progress on these questions so far...

Thank you for your attention!