

Abstract: One of the key advantages of a frame is its redundancy. Provided we have a control on the frame bounds, this redundancy allows to achieve robust reconstruction of a signal from its frame coefficients that are corrupted by noise, rounding error, or erasures. We discuss Gabor frames (g, Λ) with generic frame set Λ and random window g and show that, with high probability, such frames have frame bounds similar to the frame bounds of randomly generated frames with independent frame vectors.

Introduction

A finite set of vectors $\Phi = \{\varphi_j\}_{j=1}^N \subset \mathbb{C}^M$ is called a **frame** with **frame bounds** $0 < A \leq B$ if, for any $x \in \mathbb{C}^M$,

$$A\|x\|_2^2 \leq \sum_{j=1}^N |\langle x, \varphi_j \rangle|^2 \leq B\|x\|_2^2.$$

We identify a frame Φ with its **synthesis matrix** Φ , having the frame vectors φ_j as its columns. Its adjoint Φ^* is called the **analysis matrix** of the frame Φ .

The optimal lower and upper frame bounds are given by

$$A = \min_{x \in \mathbb{S}^{M-1}} \sum_{j=1}^N |\langle x, \varphi_j \rangle|^2 = \sigma_{\min}^2(\Phi^*), \quad B = \max_{x \in \mathbb{S}^{M-1}} \sum_{j=1}^N |\langle x, \varphi_j \rangle|^2 = \sigma_{\max}^2(\Phi^*).$$

Problem: Given noisy measurements $c = \Phi^*x + \delta \in \mathbb{C}^N$, reconstruct x .

We construct an estimate of x using standard dual frame:

$$\tilde{x} = (\Phi\Phi^*)^{-1}\Phi c = x + (\Phi\Phi^*)^{-1}\Phi\delta.$$

Then the reconstruction error is

$$\|\tilde{x} - x\|_2^2 \leq \|(\Phi\Phi^*)^{-1}\Phi\|_2^2 \|\delta\|_2^2 = \frac{\|\delta\|_2^2}{\sigma_{\min}^2(\Phi^*)}.$$

Moreover,

$$\frac{\|\tilde{x} - x\|_2}{\|x\|_2} \leq \frac{\text{Cond}(\Phi^*)}{\text{SNR}},$$

where, $\text{Cond}(\Phi^*) = \frac{\sigma_{\max}(\Phi^*)}{\sigma_{\min}(\Phi^*)} = \frac{\sqrt{B}}{\sqrt{A}}$ is the condition number of Φ , and $\text{SNR} = \frac{\|\Phi^*x\|_2}{\|\delta\|_2}$ is the signal to noise ratio.

Goal: Bound $\sigma_{\max}(\Phi^*)$, $\sigma_{\min}(\Phi^*)$ to ensure robust signal reconstruction.

Extreme singular values are sufficiently well-studied for **random matrices with independent entries** [1], [3]. The case of **structured random matrices** corresponding to application relevant frames, such as Gabor frames, is not yet fully studied [2].

Definition of Gabor frames

- 1 **Translation** (or **time shift**) by $k \in \mathbb{Z}_M$, is given by $T_k x = (x(m-k))_{m \in \mathbb{Z}_M}$.
- 2 **Modulation** (or **frequency shift**) by $\ell \in \mathbb{Z}_M$ is given by $M_\ell x = (e^{2\pi i \ell m / M} x(m))_{m \in \mathbb{Z}_M}$.
- 3 The superposition $\pi(k, \ell) = M_\ell T_k$ of translation by k and modulation by ℓ is a **time-frequency shift operator**.

For $g \in \mathbb{C}^M \setminus \{0\}$ and $\Lambda \subset \mathbb{Z}_M \times \mathbb{Z}_M$, the set of vectors

$$(g, \Lambda) = \{\pi(k, \ell)g\}_{(k, \ell) \in \Lambda}$$

is called the **Gabor system** generated by the **window** g and the set Λ . A Gabor system which spans \mathbb{C}^M is a frame and is referred to as a **Gabor frame**.

Main results [4]

We show that for any $\epsilon \in (0, 1)$, a generic subframe (g, Λ) of $(g, \mathbb{Z}_M \times \mathbb{Z}_M)$ with $|\Lambda| = O(M^{1+\epsilon} \log M)$ is has a well-conditioned analysis matrix with high probability.

More precisely, for **structured** Λ we have

Proposition 1. Let (g, Λ) be a Gabor system with $\Lambda = F \times \mathbb{Z}_M$, $F \subset \mathbb{Z}_M$, $F \neq \emptyset$, and $g \in \mathbb{C}^M$. Then (g, Λ) is a frame if and only if $\min_{m \in \mathbb{Z}_M} \{\|g_{F_m}\|_2\} \neq 0$, where g_{F_m} is the restriction of g to the set of coefficients $F_m = \{m-k\}_{k \in F} \subset \mathbb{Z}_M$.

Moreover, in this case the optimal lower and upper frame bounds for (g, Λ) are $A = M \min_{m \in \mathbb{Z}_M} \{\|g_{F_m}\|_2^2\}$ and $B = M \max_{m \in \mathbb{Z}_M} \{\|g_{F_m}\|_2^2\}$, respectively.

Note: an analogous result is true for the the case when $\Lambda = \mathbb{Z}_M \times F$, for some $F \subset \mathbb{Z}_M$.

In the case of a **generic** frame set Λ , we have

Theorem 2. Let g be a Steinhaus window and consider a Gabor system (g, Λ) .

- 1 For any $\Lambda \subset \mathbb{Z}_M \times \mathbb{Z}_M$ and $\epsilon \in (0, 1)$, with probability at least $1 - \epsilon$,

$$\sigma_{\max}^2(\Phi_\Lambda^*) \leq \frac{|\Lambda|}{M} + \sqrt{\frac{|\Lambda|}{\epsilon} \left(1 - \frac{|\Lambda|}{M^2}\right)}.$$

- 2 Let $\epsilon \in (0, \frac{1}{2}]$ and $C > 0$ a sufficiently large constant. Let $\Lambda \subset \mathbb{Z}_M \times \mathbb{Z}_M$ be a random set, constructed so that $\mathbf{1}_{\{(k, \ell) \in \Lambda\}} \sim \text{i.i.d. Bernoulli}(\tau)$ for all $(k, \ell) \in \mathbb{Z}_M \times \mathbb{Z}_M$, where $\tau = \frac{C \log M}{M^{1-\epsilon}}$. Then, with high probability,

$$\frac{|\Lambda|}{M}(1 - \delta) \leq \sigma_{\min}^2(\Phi_\Lambda^*) \leq \sigma_{\max}^2(\Phi_\Lambda^*) \leq \frac{|\Lambda|}{M}(1 + \delta).$$

Note: the bound in Theorem 2.1 is tight for a **full Gabor frame** with $\Lambda = \mathbb{Z}_M \times \mathbb{Z}_M$. In the case when $|\Lambda| = \alpha M^2$, for some $\alpha \in (0, 1)$, the proven bound gives

$\sigma_{\max}^2(\Phi_\Lambda^*) \leq \left(\alpha + \sqrt{\frac{\alpha(1-\alpha)}{\epsilon}}\right) M$ with probability at least $1 - \epsilon$, which is similar to the bound for matrices with independent entries obtained in [1].

Idea of the proof

Let Λ be a general subset of $\mathbb{Z}_M \times \mathbb{Z}_M$. Fix any $m \in \mathbb{N}$ and consider the matrix $H = \Phi_\Lambda \Phi_\Lambda^* - \frac{|\Lambda|}{M} I_M$. Following the idea of [2], we obtain that

$$\begin{aligned} \mathbb{P} \left\{ \frac{|\Lambda|}{M}(1 - \delta) \leq \sigma_{\min}^2(\Phi_\Lambda^*) \leq \sigma_{\max}^2(\Phi_\Lambda^*) \leq \frac{|\Lambda|}{M}(1 + \delta) \right\} &= \mathbb{P} \left\{ \|H\|_2 \leq \frac{|\Lambda|}{M} \delta \right\} \\ &= \mathbb{P} \left\{ \|H\|_2^{2m} > \frac{|\Lambda|^{2m}}{M^{2m}} \delta^{2m} \right\} \leq \frac{M^{2m}}{|\Lambda|^{2m}} \delta^{-2m} \mathbb{E}(\|H^m\|_2^2) \leq \frac{M^{2m}}{|\Lambda|^{2m}} \delta^{-2m} \mathbb{E}(\text{Tr } H^{2m}). \end{aligned}$$

One can further show that [4]

$$\mathbb{E}(\text{Tr } H^m) = \sum_{\substack{j_1, j_2, \dots, j_m \in \mathbb{Z}_M, \\ j_1 \neq j_2 \neq \dots \neq j_m \neq j_1 \\ (k_1, \ell_1), \dots, (k_m, \ell_m) \in \Lambda}} e^{\frac{2\pi i}{M} \sum_{t=1}^m \ell_t (j_t - j_{t+1})} E_{j_1, \dots, j_m, k_1, \dots, k_m},$$

where

$$E_{j_1, \dots, j_m, k_1, \dots, k_m} = \begin{cases} \frac{1}{M^m}, & \text{if } \exists \alpha \in S_m, \text{ s.t. } j_t - k_t = j_{\alpha(t)} - k_{\alpha(t)-1}, \forall t \in \{1, \dots, m\}; \\ 0, & \text{otherwise.} \end{cases}$$

This translates the problem of bounding the singular values into a **combinatorial problem**.

Numerical results

- Steinhaus window g , that is, $g(m) = \frac{1}{\sqrt{M}} e^{2\pi i y_m}$ and $y_m \sim \text{i.i.d. Unif}(0, 1)$.
- Λ is chosen at random as described in Theorem 2, with $\tau = \frac{C}{M}$.

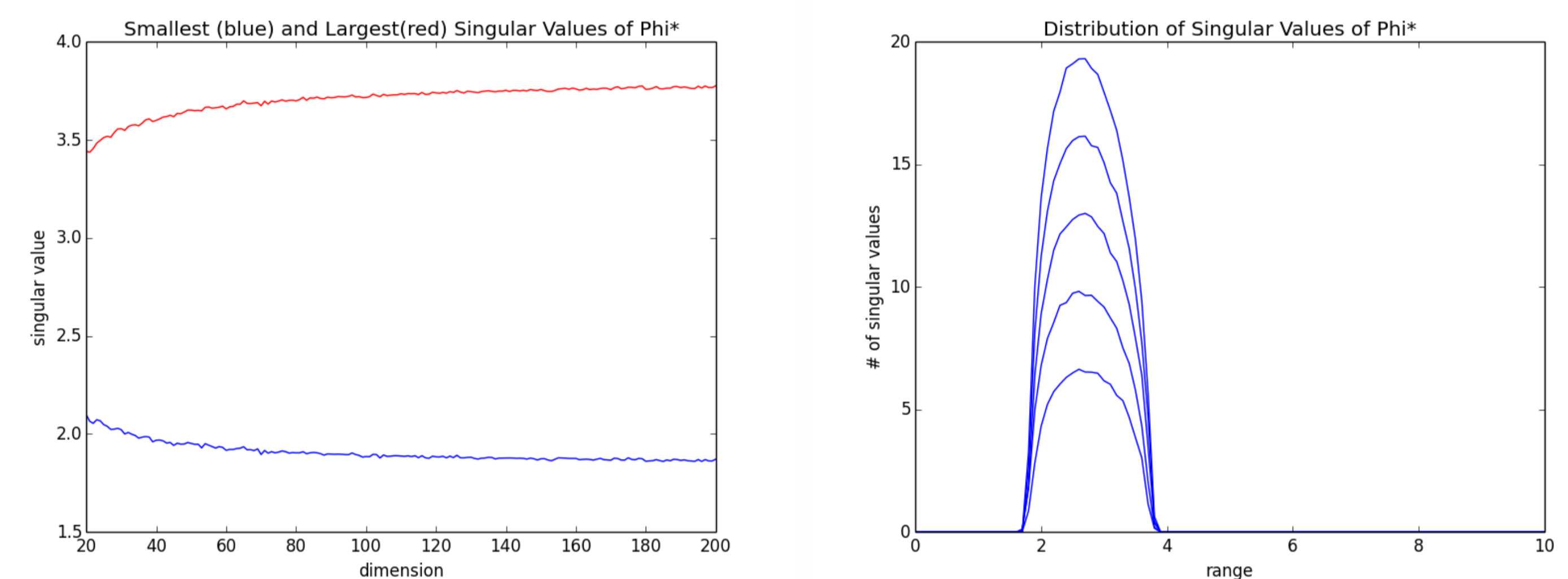


Figure 1 – Left: the dependence of the extreme singular values of the analysis matrix Φ_Λ^* of a Gabor frame (g, Λ) on the ambient dimension M ; Right: the distribution of the singular values of Φ_Λ^* for the dimensions $M = 100, 150, 200, 250, 300$.

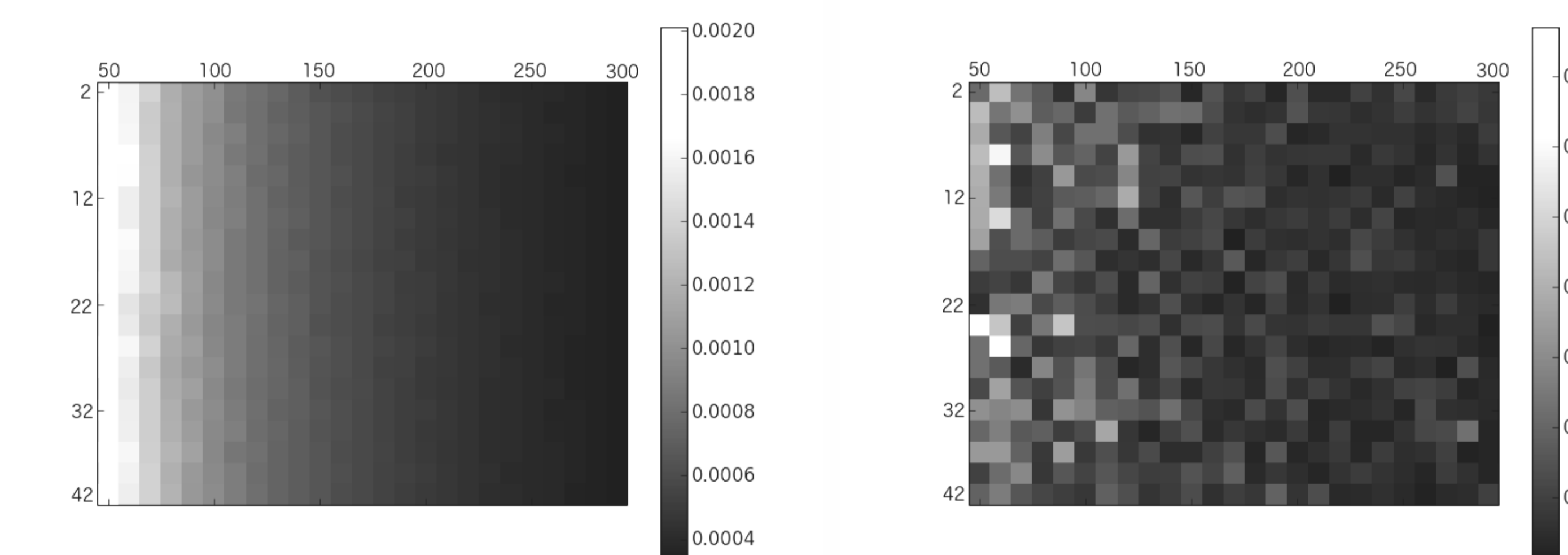


Figure 2 – The dependence of the numerically estimated trace expectation $\frac{M^{2m}}{|\Lambda|^{2m}} \delta^{-2m} \mathbb{E}(\text{Tr } H^{2m})$ on the ambient dimension M (horizontal axis) and the parameter C (vertical axis), for a fixed m . Left: Λ is chosen at random, as described in Theorem 2, with $\tau = \frac{C}{M}$; Right: $\Lambda = F \times \{0, 1, \dots, \lfloor \frac{M}{2} \rfloor\}$.

The obtained numerical results suggest that

- 1 In the case when random Λ is constructed as in Theorem 2 with $\tau = \frac{C}{M}$, there exist constants $0 < k < K$ not depending on M , such that all the singular values of the analysis matrix Φ_Λ^* are inside the interval $\left[\frac{k|\Lambda|}{M}, \frac{K|\Lambda|}{M}\right]$ with high probability.
- 2 Even in the worst case scenario choice of Λ normalized trace expectation decreases rapidly with the dimension. This allows to conjecture that Theorem 2 can be further generalized using the proposed method.

Conclusions and Forthcoming Research

While the presented results discuss the case of a generic, randomly generated, Λ , one of the main directions for the future research is to evaluate frame bounds of Gabor frames for all possible frame sets Λ and to investigate their dependencies on the structure of Λ .

References

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