Group-theoretic constructions of erasure-robust frames

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The Restricted Isometry Property (RIP)

Definition: Fix $K \leq M \leq N$ and let $\Phi = [\varphi_1 \cdots \varphi_N] \in \mathbb{R}^{M \times N}$.

We say Φ has the (K, δ) -**Restricted Isometry Property (RIP)** if for every *K*-element subset *K* of $\{1, \ldots, N\}$, we have

$$(1-\delta)\sum_{n\in\mathcal{K}}|y(n)|^2\leq \left\|\sum_{n\in\mathcal{K}}y(n)\varphi_n\right\|^2\leq (1+\delta)\sum_{n\in\mathcal{K}}|y(n)|^2,$$

for all $y \in \mathbb{R}^N$.

Fact: For any *K*-element subset \mathcal{K} of $\{1, \ldots, N\}$, consider the $M \times K$ submatrix $\Phi_{\mathcal{K}}$ of Φ with columns $\{\varphi_n\}_{n \in \mathcal{K}}$.

Then Φ is (\mathcal{K}, δ) -RIP if and only if the eigenvalues of $\Phi_{\mathcal{K}}^{T} \Phi_{\mathcal{K}}$ lie in $[1 - \delta, 1 + \delta]$ for all \mathcal{K} .

RIP: Il buono, il brutto, il cattivo

The Good: Candès and Tao showed that L1-minimization can be used to quickly and stably find a unique *K*-sparse solution *y* to an *underdetermined* linear system $\Phi y = z$, provided the matrix Φ is $(2K, \delta)$ -RIP and a sparse solution exists.

Moreover, they showed that with overwhelming probability, certain random matrices will be (K, δ) -RIP for $K = O(M/\log(N))$.

The Bad: All known deterministic constructions of RIP matrices are only guaranteed to be (K, δ) -RIP for $K = \mathcal{O}(M^{\frac{1}{2}+\varepsilon})$. This is known as the **square root bottleneck**.

The Ugly: Directly checking whether or not a given Φ has the RIP involves estimating the singular values of $\binom{N}{K}$ possible submatrices of Φ ; see "Certifying the Restricted Isometry Property is Hard" by Bandeira, Dobriban, Mixon and Sawin (2013).

Numerically Erasure-Robust Frames (NERFs)

Definition: Fix $M \leq K \leq N$ and let $\Phi = [\varphi_1 \cdots \varphi_N] \in \mathbb{R}^{M \times N}$.

We say $\{\varphi_n\}_{n=1}^N$ is a (K, α, β) -NERF for \mathbb{R}^M if there exists $0 < \alpha \leq \beta < \infty$ such that for every K-element subset \mathcal{K} of $\{1, \ldots, N\}$ we have $\{\varphi_n\}_{n \in \mathcal{K}}$ is a **frame** for \mathbb{R}^M with frame bounds α and β :

$$\alpha \|x\|^2 \leq \sum_{n \in \mathcal{K}} |\langle x, \varphi_n \rangle|^2 \leq \beta \|x\|^2, \quad \forall x \in \mathbb{R}^M.$$

Here we want the eigenvalues of $\Phi_{\mathcal{K}} \Phi_{\mathcal{K}}^{T}$ to lie in $[\alpha, \beta]$ for all \mathcal{K} .

Recall: For $K \leq M \leq N$, Φ has the (K, δ) -RIP if the eigenvalues of $\Phi_{\mathcal{K}}^{\mathrm{T}} \Phi_{\mathcal{K}}$ lie in $[1 - \delta, 1 + \delta]$ for all $\mathcal{K} \subseteq \{1, \ldots, N\}$, $|\mathcal{K}| = K$.

Optimal NERF Bounds

Note: For any fixed K, the optimal NERF bounds α_K and β_K for $\{\varphi_n\}_{n=1}^N$ are the extreme eigenvalues of $\Phi_K^T \Phi_K$:

$$\alpha_{\mathcal{K}} := \min_{|\mathcal{K}| = \mathcal{K}} \min_{\|x\| = 1} \sum_{n \in \mathcal{K}} |\langle x, \varphi_n \rangle|^2, \ \beta_{\mathcal{K}} := \max_{|\mathcal{K}| = \mathcal{K}} \max_{\|x\| = 1} \sum_{n \in \mathcal{K}} |\langle x, \varphi_n \rangle|^2.$$

Estimating $\alpha_{\mathcal{K}}$ and $\beta_{\mathcal{K}}$ thus seems combinatorially difficult.

Idea: Rather than find the "worst x for any \mathcal{K} ," let's instead find the "worst \mathcal{K} for any x," namely interchange the optimizations:

$$\alpha_{K} := \min_{\|x\|=1} \min_{|\mathcal{K}|=K} \sum_{n \in \mathcal{K}} |\langle x, \varphi_{n} \rangle|^{2}, \ \beta_{K} := \max_{\|x\|=1} \max_{|\mathcal{K}|=K} \sum_{n \in \mathcal{K}} |\langle x, \varphi_{n} \rangle|^{2}.$$

For a fixed x, these worst \mathcal{K} 's are found by sorting $\{|\langle x, \varphi_n \rangle|^2\}_{n=1}^N$.

Problem: There are an infinite number of *x*'s on the unit sphere.

ε -Nets

Definition: Given $\varepsilon > 0$, a sequence of unit norm vectors $\{\psi_p\}_{p=1}^P$ is an ε -net for \mathbb{S}^{M-1} (with respect to chordal distance) if for all $x \in \mathbb{S}^{M-1}$ there exists ψ_p such that $|\langle x, \psi_p \rangle|^2 \ge 1 - \varepsilon^2$.

Idea: Given a ε -net $\{\psi_p\}_{p=1}^P$ for \mathbb{S}^{M-1} , estimate the optimal NERF bounds for $\{\varphi_n\}_{n=1}^N$ as the ε -approximate NERF bounds:

$$\alpha_{K,\varepsilon} := \min_{p=1,\dots,P} \sum_{n=1}^{K} |\langle \psi_p, \varphi_{\sigma(n)} \rangle|^2,$$
$$\beta_{K,\varepsilon} := \max_{p=1,\dots,P} \sum_{n=N-K+1}^{N} |\langle \psi_p, \varphi_{\sigma(n)} \rangle|^2,$$

where σ is a *p*-dependent permutation of $\{1, \ldots, N\}$ chosen so that the values $\{|\langle \psi_p, \varphi_{\sigma(n)} \rangle|^2\}_{n=1}^N$ are nondecreasing.

Estimating NERF Bounds with ε -Nets

Theorem: [FJMP (2012)] Optimal NERF bounds α_K and β_K are estimated by ε -approximate bounds $\alpha_{K,\varepsilon}$ and $\beta_{K,\varepsilon}$ according to

$$\frac{1}{1-\varepsilon^2} \left(\alpha_{K,\varepsilon} - \frac{\varepsilon^2}{1-\varepsilon^2} \beta_{K,\varepsilon} \right) \le \alpha_K \le \alpha_{K,\varepsilon}, \quad \beta_{K,\varepsilon} \le \beta_K \le \frac{1}{1-\varepsilon^2} \beta_{K,\varepsilon}.$$

Moreover, if $\{\varphi_n\}_{n=1}^N$ is a **unit-norm tight frame (UNTF)** $(\Phi \Phi^{\mathrm{T}} = \frac{N}{M} \mathrm{I} \text{ and } \|\varphi_n\| = 1 \text{ for all } n)$ then we also have

$$\frac{1}{1-\varepsilon^2}(\alpha_{K,\varepsilon}-\varepsilon^2\frac{N}{M})\leq \alpha_K\leq \beta_K\leq \frac{N}{M}.$$

Note: For every fixed ψ_p , we compute $\{|\langle \psi_p, \varphi_n \rangle|^2\}_{n=1}^N$, and then sort these values so as to sum the K smallest and largest ones. Taking the minimum and maximum of these sums over all p yields $\alpha_{K,\varepsilon}$ and $\beta_{K,\varepsilon}$. This uses $\mathcal{O}((M + \log N)NP)$ operations overall.

Problem: Good ε -nets are enormous, e.g. $P = (1 + \frac{2}{\varepsilon})^M$.

Group Frames

Definition: Let $\mathcal{U} = \{U_q\}_{q=1}^Q$ be a finite group of $M \times M$ orthogonal matrices. We say that $\{\varphi_n\}_{n=1}^N$ is \mathcal{U} -invariant if

 $\forall q \exists$ a permutation σ of $\{1, \ldots, N\}$ s.t. $U_q \varphi_n = \pm \varphi_{\sigma(n)}, \forall n$.

Note: We focus exclusively on the $2^M M!$ -element group of signed permutation matrices that arises the symmetry group of the hypercube in \mathbb{R}^M . This group is **irreducible** meaning the orbit of any unit norm vector under its action is a UNTF.

Example: When M = 4, there are $2^{4}4! = (16)(24) = 384$ distinct 4×4 signed permutation matrices. The following 4×12 UNTF is invariant under the action of this group:

Estimating NERF Bounds with Group-Generated ε -Nets

Idea: If $\{\varphi_n\}_{n=1}^N$ is \mathcal{U} -invariant, then $\forall q$, $\{|\langle U_q\psi_r, \varphi_n\rangle|^2\}_{n=1}^N$ has the same K smallest and largest values as $\{|\langle \psi_r, \varphi_n\rangle|^2\}_{n=1}^N$.

Theorem: [FJMP (2012)] Let \mathcal{U} be a finite group of $M \times M$ orthogonal matrices and let $\{\varphi_n\}_{n=1}^N$ be \mathcal{U} -invariant. Choosing $\{\psi_r\}_{r=1}^R \subseteq \mathbb{S}^{M-1}$ such that $\{U_q\psi_r\}_{q=1, r=1}^Q$ is an ε -net for \mathbb{S}^{M-1} , the corresponding ε -approximate NERF bounds are:

$$\alpha_{K,\varepsilon} = \min_{r=1,\dots,R} \sum_{n=1}^{K} |\langle \psi_r, \varphi_{\sigma(n)} \rangle|^2,$$
$$\beta_{K,\varepsilon} = \max_{r=1,\dots,R} \sum_{n=N-K+1}^{N} |\langle \psi_r, \varphi_{\sigma(n)} \rangle|^2,$$

where σ is chosen so that $\{|\langle \psi_r, \varphi_{\sigma(n)} \rangle|^2\}_{n=1}^N$ is nondecreasing.

An ε -Net for Nonnegative, Nonincreasing Vectors

Note: For any $x \in \mathbb{S}^{M-1}$, there exists a signed permutation U_q such that $U_q x$ is nonnegative and nonincreasing, namely such that

$$U_q x \in \mathbb{S}_{\mathrm{nn}}^{M-1} := \{x \in \mathbb{S}^{M-1} : 0 \le x(1) \le \cdots \le x(M)\}.$$

Lemma: [FJMP (2012)] Let $\{\psi_r\}_{r=1}^R \subseteq \mathbb{S}_{nn}^{M-1}$ and let $\{U_q\}_{q=1}^Q$ be the group all $M \times M$ signed permutations. Then $\{U_q\psi_r\}_{q=1, r=1}^Q$ is an ε -net for \mathbb{S}^{M-1} if and only if $\{\psi_r\}_{r=1}^R$ is an ε -net for \mathbb{S}_{nn}^{M-1} .

Note: When combined with the previous result, this means that in order to estimate the NERF bounds of a \mathcal{U} -invariant frame, we only need to compute $\{|\langle \psi_r, \varphi_n \rangle|^2\}_{n=1}^N$ at every point ψ_r of an ε -net for \mathbb{S}_{nn}^{M-1} instead of at every point of an ε -net for \mathbb{S}^{M-1} .

The surface area of \mathbb{S}_{nn}^{M-1} is that of \mathbb{S}^{M-1} divided by $2^M M!$.

Vector Quantization by "Rounding Up"

Lemma: [FJMP (2012)] For any positive integer M and $\varepsilon > 0$, let $\delta = [M(L-1)]^{-\frac{1}{2L}}$ and take any $L \ge 2$ such that

$$(L-1)(1-\varepsilon^2)^L \leq \frac{1}{M} \left(\frac{L-1}{L}\right)^L.$$

Then for any $x\in \mathbb{S}_{\mathrm{nn}}^{M-1}$, the step function $\psi_x=\hat{\psi}_x/\|\hat{\psi}_x\|$,

$$\hat{\psi}_{x}(m) := \begin{cases} \delta', & \delta^{l+1} < x(m) \leq \delta', \\ \delta^{L-1}, & 0 \leq x(m) \leq \delta^{L-1} \end{cases}$$

satisfies $\langle x, \psi_x \rangle > \sqrt{1 - \varepsilon^2}$.

Note: The set of all such ψ_x 's forms an ε -net for \mathbb{S}_{nn}^{M-1} . Since each ψ_x arises from a unique nonincreasing $\{1, \ldots, L\}$ -valued function over $\{1, \ldots, M\}$, "stars and bars" reveals the number of elements in this ε -net to be at most $\binom{M+L-1}{L-1} \leq C_1 M^{C_2(\varepsilon) \log M}$.

Main Result

Theorem: [FJMP (2012)] Let $\{\varphi_n\}_{n=1}^N$ be a UNTF for \mathbb{R}^M which is invariant under signed permutations. For any $\varepsilon > 0$, take δ and L as in the previous lemma and construct $\{\psi_r\}_{r=1}^R$ by normalizing all $\{\delta'\}_{l=0}^{L-1}$ -valued nondecreasing step functions.

Then for any $M \leq K \leq N$, the optimal NERF bounds α_K and β_K of $\{\varphi_n\}_{n=1}^N$ satisfy the estimates

$$\begin{aligned} \frac{1}{1-\varepsilon^2} \Big(\alpha_{K,\varepsilon} - \varepsilon^2 \min\left\{\frac{N}{M}, \frac{1}{1-\varepsilon^2} \beta_{K,\varepsilon} \right\} \Big) &\leq \alpha_K \leq \alpha_{K,\varepsilon}, \\ \beta_{K,\varepsilon} &\leq \beta_K \leq \min\left\{\frac{N}{M}, \frac{1}{1-\varepsilon^2} \beta_{K,\varepsilon} \right\}, \end{aligned}$$

where $\alpha_{{\cal K},\varepsilon}$ and $\beta_{{\cal K},\varepsilon}$ are found by the following process:

For any $r = 1, \ldots, R$, let $\alpha_{K,\varepsilon,r}$ and $\beta_{K,\varepsilon,r}$ be the sums of the K smallest and largest values of $\{|\langle \psi_r, \varphi_n \rangle|^2\}_{n=1}^N$, respectively.

Let
$$\alpha_{K,\varepsilon} = \min_{r} \alpha_{K,\varepsilon,r}$$
 and $\beta_{K,\varepsilon} = \max_{r} \beta_{K,\varepsilon,r}$.

Numerical Example: M = 4, N = 12

Size of ε -net:	ε^2	L	$\binom{M+L-1}{L-1}$	$R_{ m improved}$
	2^{-1}	6	126	45
	2^{-2}	19	7315	1107
	2^{-3}	47	230300	15916
	2^{-4}	110	6438740	202628
	2^{-5}	249	164059875	2366922
	-			D

ε -approximate lower NERF bounds $\alpha_{K,\varepsilon}$:

	$\varepsilon^2 \setminus K$	1	2	3	4	5	6	7	8	9	10	11	12	_	
	2^{-1}	0.00	0.00	0.00	0.00	0.00	0.00	0.38	0.72	1.00	1.58	2.10	3.00		
	2^{-2}	0.00	0.00	0.00	0.00	0.00	0.00	0.38	0.71	1.00	1.52	2.03	3.00		
	2^{-3}	0.00	0.00	0.00	0.00	0.00	0.00	0.38	0.71	1.00	1.50	2.01	3.00		
	2^{-4}	0.00	0.00	0.00	0.00	0.00	0.00	0.38	0.71	1.00	1.50	2.00	3.00		
	2^{-5}	0.00	0.00	0.00	0.00	0.00	0.00	0.38	0.71	1.00	1.50	2.00	3.00		
	α_K	0.00	0.00	0.00	0.00	0.00	0.00	0.38	0.71	1.00	1.50	2.00	3.00	_	
$\frac{1}{1-\varepsilon^2}$	$\left(\alpha_{K,i}\right)$	$\varepsilon - \varepsilon$	² miı	$n\left\{\frac{N}{M},\right.$	$rac{1}{1-arepsilon^2}$	$\beta_{K,\varepsilon}$	·):								
~															
$\varepsilon^2 \setminus K$	(1	2	3	4	!	5	6	7	8		9	10	11	12
$\varepsilon^2 \setminus K$ 2^{-1}	-3.0	1 00 —3	2	3	4	-3.0	5 0 —	6 3.00	7	8	-0.	9 99 0	10	11 21	3.00
$\frac{\varepsilon^2 \setminus K}{2^{-1}}$	-3.0 -1.0	1 00 -3 00 -1	2 3.00 1.00	3 -3.00 -1.00	4 -3.00 -1.00	-3.0 -1.0	5 D — D —	6 3.00 1.00	7 -2.23 -0.49	8 -1.54 -0.04	-0. 0.	9 99 0 33 1	10 0.16 1	11 .21 .71	12 3.00 3.00
$\frac{\varepsilon^2 \setminus K}{2^{-1}}$ $\frac{2^{-2}}{2^{-3}}$	-3.0 -1.0 -0.4	$ \begin{array}{c} 1 \\ 00 \\ -3 \\ 00 \\ -2 \\ -0 \end{array} $	2 3.00 1.00 0.42	3 -3.00 -1.00 -0.42	4 -3.00 -1.00 -0.42	-3.0 -1.0 -0.4	5 0 — 0 — 2 —	6 3.00 1.00 0.42	7 -2.23 -0.49 0.00	8 -1.54 -0.04 0.39	-0. 0. 0.	9 99 0 33 1 71 1	10 0.16 1 0.02 1 0.29 1	11 21 71 87	3.00 3.00 3.00
$\frac{\varepsilon^2 \setminus K}{2^{-1}}$ $\frac{2^{-2}}{2^{-3}}$ $\frac{2^{-3}}{2^{-4}}$	-3.0 -1.0 -0.4 -0.2	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	2 3.00 1.00 0.42 0.20	3 -3.00 -1.00 -0.42 -0.20	4 -3.00 -1.00 -0.42 -0.20	-3.0 -1.0 -0.4 -0.2	5 0 — 2 — 0 —	6 3.00 1.00 0.42 0.20	7 -2.23 -0.49 0.00 0.20	8 -1.54 -0.04 0.39 0.56	-0. 0. 0. 0.	9 99 0 33 1 71 1 86 1	10 02 1 29 1 40 1	11 1.21 1.71 1.87 1.93	3.00 3.00 3.00 3.00 3.00
$\frac{\varepsilon^{2} \setminus K}{2^{-1}}$ $\frac{2^{-2}}{2^{-3}}$ $\frac{2^{-4}}{2^{-5}}$	-3.0 -1.0 -0.4 -0.2 -0.0	$ \begin{array}{cccccccccccccccccccccccccccccccccccc$	2 3.00 1.00 0.42 0.20 0.09	3 -3.00 -1.00 -0.42 -0.20 -0.09	4 -3.00 -1.00 -0.42 -0.20 -0.09	-3.00 -1.00 -0.42 -0.20	5 0 — 2 — 2 — 9 —	6 3.00 1.00 0.42 0.20 0.09	7 -2.23 -0.49 0.00 0.20 0.29	8 -1.54 -0.04 0.39 0.56 0.64	-0. 0. 0. 0.	9 99 0 33 1 71 1 86 1 93 1	10 0.16 0.02 0.29 0.40 0.45	11 1.21 1.71 1.87 1.93 1.96	3.00 3.00 3.00 3.00 3.00

More Numerical Examples

Example: Let M = 6 and let $\{\varphi_n\}_{n=1}^N$ be the 80 signed permutations of $\varphi = [1 \ 1 \ 1 \ 0 \ 0 \ 0]^T$ which are distinct modulo negation. Taking $\varepsilon = \frac{1}{2}$, our MATLAB code took around 8.84 seconds to show that any 61 of these 80 frame elements span \mathbb{R}^6 . Obtaining this same fact directly involves forming each of the $\binom{80}{61} \approx 1.16 \times 10^{18}$ such submatrices.

Example: Let M = 8 and let $\{\varphi\}_{n=1}^{N}$ be the 560 distinct signed permutations of $\varphi = [1 \ 1 \ 1 \ 1 \ 0 \ 0 \ 0]^{\mathrm{T}}$. Taking $\varepsilon = \frac{1}{2}$, our MATLAB code took around three minutes to show that any 399 of these 560 frame elements span \mathbb{R}^{9} . Note $\binom{560}{399} \approx 2.94 \times 10^{144}$.

Example: Let M = 10 and let $\{\varphi\}_{n=1}^{N}$ be the 4032 distinct signed permutations of $\varphi = [1 \ 1 \ 1 \ 1 \ 1 \ 0 \ 0 \ 0 \ 0]^{\mathrm{T}}$. Taking $\varepsilon = \frac{1}{2}$, our MATLAB code took around 77 minutes to show that any 2883 of these 4032 frame elements span \mathbb{R}^{10} . Note $\binom{4032}{2883} \approx 3.65 \times 10^{1044}$.

Conclusions and Future Work

Conclusions:

- ► Using explicit *ε*-nets reduces the problem of numerically estimating the optimal NERF bounds of any given finite frame from being exponential in *N* to being exponential in *M*.
- Exploiting symmetry can further reduce this problem to being subexponential in *M* provided the frame is group-invariant.
- Even after these speedups, estimating NERF bounds still involves a large amount of computation.

Future Work:

- Methods for constructing smaller ε -nets for \mathbb{S}_{nn}^{M-1} ?
- Apply these techniques to the deterministic RIP problem?