

Deterministic and Stochastic Bounds in the Phase Retrieval Problem

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January 13, 2016
Applied Mathematics Seminar – Stanford University

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"This material is based upon work supported by the National Science Foundation under Grant No. DMS-1413249. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation."

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Problem Formulation

The phase retrieval problem

- Hilbert space $H = \mathbb{C}^n$, $\hat{H} = H/T^1$, frame $\mathcal{F} = \{f_1, \dots, f_m\} \subset \mathbb{C}^n$ and

$$\alpha : \hat{H} \rightarrow \mathbb{R}^m, \quad \alpha(x) = (|\langle x, f_k \rangle|)_{1 \leq k \leq m}.$$

$$\beta : \hat{H} \rightarrow \mathbb{R}^m, \quad \beta(x) = (|\langle x, f_k \rangle|^2)_{1 \leq k \leq m}.$$

The frame is said *phase retrievable* (or that it gives phase retrieval) if α (or β) is injective.

- The general *phase retrieval problem* a.k.a. *phaseless reconstruction*:
Decide when a given frame is phase retrievable, and, if so, find an algorithm to recover x from $y = \alpha(x)$ (or from $y = \beta(x)$) up to a global phase factor.

Problem Formulation

Lipschitz Reconstruction

Assume $\mathcal{F}$ is phase retrievable.

Our Problems Today:

- 1 Are the nonlinear maps α, β bi-Lipschitz with respect to appropriate metrics?
- 2 Do they admit left inverses that are globally Lipschitz?
- 3 What are the Lipschitz constants? What is the structure of local Lipschitz bounds?
- 4 What is the average performance of any reconstruction scheme (Cramer-Rao Lower Bounds)?

1-3: Worst Case Performance

4: Average Case Performance

Metric Space Structures

Topological Structures

Let $H = \mathbb{C}^n$. The quotient space $\hat{H} = \mathbb{C}^n / T^1$, with classes induced by $x \sim y$ if there is real φ with $x = e^{i\varphi}y$.

Topologically:

$$\hat{\mathbb{C}}^n = \{0\} \cup \left((0, \infty) \times \mathbb{CP}^{n-1} \right)$$

with

$$\mathring{\mathbb{C}}^n = \hat{\mathbb{C}}^n \setminus \{0\} = (0, \infty) \times \mathbb{CP}^{n-1}$$

a real analytic manifold of real dimension $2n - 1$.

Another embedding is into the space of symmetric matrices $Sym(\mathbb{C}^n)$.

Specifically let

$$\mathcal{S}^{p,q}(H) = \{T \in Sym(H) \text{ , } T \text{ has at most } p \text{ pos.eigs. and } q \text{ neg.eigs}\}$$

Then:

$$\kappa_\beta : \hat{H} \rightarrow \mathcal{S}^{1,0} \text{ , } \hat{x} \mapsto xx^* \text{ , is an embedding.}$$

Metric Space Structures

The matrix norm-induced metric structure

Fix $1 \leq p \leq \infty$. The *matrix-norm induced distance*

$$d_p : \hat{H} \times \hat{H} \rightarrow \mathbb{R}, \quad d_p(\hat{x}, \hat{y}) = \|xx^* - yy^*\|_p$$

with the p -norm of the singular values. In the case $p = 2$ we obtain

$$d_2(x, y) = \sqrt{\|x\|^4 + \|y\|^4 - 2|\langle x, y \rangle|^2}$$

Lemma (BZ15)

- ① $(d_p)_{1 \leq p \leq \infty}$ are equivalent metrics and the identity map $i : (\hat{H}, d_p) \rightarrow (\hat{H}, d_q)$, $i(x) = x$ has Lipschitz constant

$$\text{Lip}_{p,q,n}^d = \max(1, 2^{\frac{1}{q} - \frac{1}{p}}).$$

- ② The metric space $(\hat{H}, d_p)$ is isometrically isomorphic to $\mathcal{S}^{1,0}$ endowed with the p -norm via $\kappa_\beta : \hat{H} \rightarrow \mathcal{S}^{1,0}$, $x \mapsto \kappa_\beta(x) = xx^*$.

Metric Space Structures

The natural metric structure

Fix $1 \leq p \leq \infty$. The *natural metric*

$$D_p : \hat{H} \times \hat{H} \rightarrow \mathbb{R}, \quad D_p(\hat{x}, \hat{y}) = \min_{\varphi} \|x - e^{i\varphi} y\|_p$$

with the usual p -norm on $\mathbb{C}^n$. In the case $p = 2$ we obtain

$$D_2(\hat{x}, \hat{y}) = \sqrt{\|x\|^2 + \|y\|^2 - 2|\langle x, y \rangle|}$$

Lemma (BZ15)

- ① $(D_p)_{1 \leq p \leq \infty}$ are equivalent metrics and the identity map $i : (\hat{H}, D_p) \rightarrow (\hat{H}, D_q)$, $i(x) = x$ has Lipschitz constant

$$\text{Lip}_{p,q,n}^D = \max(1, n^{\frac{1}{q} - \frac{1}{p}}).$$

- ② The metric space $(\hat{H}, D_2)$ is Lipschitz isomorphic to $\mathcal{S}^{1,0}$ endowed with the 2-norm via $\kappa_\alpha : \hat{H} \rightarrow \mathcal{S}^{1,0}$, $x \mapsto \kappa_\alpha(x) = \frac{1}{\|x\|} x x^*$.

Metric Space Structures

Distinct Structures

Two different structures: topologically equivalent, BUT the metrics are NOT equivalent:

Lemma (BZ15)

The identity map $i : (\hat{H}, D_p) \rightarrow (\hat{H}, d_p)$, $i(x) = x$ is continuous but it is not Lipschitz continuous. Likewise, the identity map $i : (\hat{H}, d_p) \rightarrow (\hat{H}, D_p)$, $i(x) = x$ is continuous but it is not Lipschitz continuous. Hence the induced topologies on $(\hat{H}, D_p)$ and $(\hat{H}, d_p)$ are the same, but the corresponding metrics are not Lipschitz equivalent.

Main Results

Lipschitz inversion: α

Theorem (BZ15)

Assume $\mathcal{F}$ is a phase retrievable frame for H . Then:

- 1 The map $\alpha : (\hat{H}, D_2) \rightarrow (\mathbb{R}^m, \|\cdot\|_2)$ is bi-Lipschitz. Let $\sqrt{A_0}, \sqrt{B_0}$ denote its Lipschitz constants: for every $x, y \in H$:

$$A_0 \min_{\varphi} \|x - e^{i\varphi} y\|_2^2 \leq \sum_{k=1}^m |\langle x, f_k \rangle| - |\langle y, f_k \rangle| \leq B_0 \min_{\varphi} \|x - e^{i\varphi} y\|_2^2.$$

- 2 There is a Lipschitz map $\omega : (\mathbb{R}^m, \|\cdot\|_2) \rightarrow (\hat{H}, D_2)$ so that: (i) $\omega(\alpha(x)) = x$ for every $x \in \hat{H}$, and (ii) its Lipschitz constant is $Lip(\omega) \leq \frac{4+3\sqrt{2}}{\sqrt{A_0}} = \frac{8.24}{\sqrt{A_0}}$.

Main Results

Lipschitz inversion: β

Theorem (BZ15)

Assume $\mathcal{F}$ is a phase retrievable frame for H . Then:

- 1 The map $\beta : (\hat{H}, d_1) \rightarrow (\mathbb{R}^m, \|\cdot\|_2)$ is bi-Lipschitz. Let $\sqrt{a_0}, \sqrt{b_0}$ denote its Lipschitz constants: for every $x, y \in H$:

$$a_0 \|xx^* - yy^*\|_1^2 \leq \sum_{k=1}^m \left| |\langle x, f_k \rangle|^2 - |\langle y, f_k \rangle|^2 |^2 \leq b_0 \|xx^* - yy^*\|_1^2.$$

- 2 There is a Lipschitz map $\psi : (\mathbb{R}^m, \|\cdot\|_2) \rightarrow (\hat{H}, d_1)$ so that: (i) $\psi(\beta(x)) = x$ for every $x \in \hat{H}$, and (ii) its Lipschitz constant is $Lip(\psi) \leq \frac{4+3\sqrt{2}}{\sqrt{a_0}} = \frac{8.24}{\sqrt{a_0}}$.

Main Results

Statistical models

- A general noisy measurement process is given by:

$$y_k = |\langle x, f_k \rangle + \mu_k|^p + \nu_k, \quad 1 \leq k \leq m,$$

where $(\mu_k)_k, (\nu_k)_k$ are two noise processes.

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- AWGN Model: $\mu_k = 0$, $p = 2$ and $\nu_k \sim \mathbb{N}(0, \sigma^2)$ i.i.d.

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- Non-AWGN Model: $\mu_k \sim \mathbb{CN}(0, \rho^2)$, i.i.d. and $\nu_k = 0$,

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$$y_k = |\langle x, f_k \rangle + \mu_k|^p, \quad 1 \leq k \leq m.$$

Want:

- 1) Fisher Information Matrix $\mathbb{I} = \mathbb{E}[(\nabla_x \log p(y; x))(\nabla_x \log p(y; x))^*]$.
- 2) Cramer-Rao Lower Bounds for unbiased estimators.

Main Results

Fisher Information Matrix

$$\mathbb{I}^{AWGN, real}(x) = \frac{4}{\sigma^2} \sum_{k=1}^m |\langle x, f_k \rangle|^2 f_k f_k^T = \frac{4}{\sigma^2} \sum_{k=1}^m (f_k f_k^T) x x^T (f_k f_k^T) \quad [\text{Bal12}].$$

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$$\mathbb{I}^{AWGN,cplx}(x) = \frac{4}{\sigma^2} \sum_{k=1}^m \Phi_k \xi \xi^* \Phi_k \quad [\text{Bal13}, \text{BCMN13}].$$

Main Results

Fisher Information Matrix

$$\mathbb{I}^{AWGN,real}(x) = \frac{4}{\sigma^2} \sum_{k=1}^m |\langle x, f_k \rangle|^2 f_k f_k^T = \frac{4}{\sigma^2} \sum_{k=1}^m (f_k f_k^T)_{xx^T} (f_k f_k^T) \quad [\text{Bal12}].$$

$$\mathbb{I}^{AWGN,cplx}(x) = \frac{4}{\sigma^2} \sum_{k=1}^m \Phi_k \xi \xi^* \Phi_k \quad [\text{Bal13}, \text{BCMN13}].$$

$$\begin{aligned} \mathbb{I}^{nonAWGN,cplx}(x) &= \frac{4}{\rho^4} \sum_{k=1}^m \left(G_1 \left(\frac{\langle \Phi_k \xi, \xi \rangle}{\rho^2} \right) - 1 \right) \Phi_k \xi \xi^* \Phi_k \\ &= \frac{4}{\rho^2} \sum_{k=1}^m G_2 \left(\frac{\langle \Phi_k \xi, \xi \rangle}{\rho^2} \right) \frac{1}{\langle \Phi_k \xi, \xi \rangle} \Phi_k \xi \xi^* \Phi_k \quad [\text{Bal15}]. \end{aligned}$$

where

$$G_1(a) = \frac{e^{-a}}{8a^3} \int_0^\infty \frac{l_1^2(t)}{l_0(t)} t^3 e^{-\frac{t^2}{4a}} dt, \quad G_2(a) = a(G_1(a) - 1).$$

Main Results

AWGN vs. non-AWGN: Comparisons and Identifiability

Let B be the frame upper bound.

Lemma

$$\frac{\sigma^2}{\rho^4} \left(G_1\left(\frac{B\|x\|^2}{\rho^2}\right) - 1 \right) \mathbb{I}^{AWGN, cplx}(x) \leq \mathbb{I}^{nonAWGN, cplx}(x) \leq \frac{\sigma^2}{\rho^4} \mathbb{I}^{AWGN, cplx}(x)$$

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Lemma

$$\frac{\sigma^2}{\rho^4} \left(G_1\left(\frac{B\|x\|^2}{\rho^2}\right) - 1 \right) \mathbb{I}^{AWGN, cplx}(x) \leq \mathbb{I}^{nonAWGN, cplx}(x) \leq \frac{\sigma^2}{\rho^4} \mathbb{I}^{AWGN, cplx}(x)$$

Theorem

The following are equivalent:

- ① The frame $\mathcal{F}$ is phase retrievable;
- ② For every $0 \neq x \in \mathbb{C}^n$, $\text{rank}(\mathbb{I}^{nonAWGN, cplx}(x)) = 2n - 1$;
- ③ For every $0 \neq x \in \mathbb{C}^n$, $\text{rank}(\mathbb{I}^{AWGN, cplx}(x)) = 2n - 1$;

Main Results

The Cramer-Rao Lower Bound

Fix $z_0 \in \mathbb{C}^n$, $\|z_0\| = 1$, let $\zeta_0 = [\text{real}(z_0) \ \text{imag}(z_0)]^T$ and set

$$\Omega_{z_0} = \{\xi \in \mathbb{R}^{2n} \ , \ \langle \xi, \zeta_0 \rangle \geq 0, \langle \xi, J\zeta_0 \rangle = 0\}.$$

Let $\Pi_{z_0} = 1 - J\zeta_0\zeta_0^*J^*$ with J the symplectic form matrix.

Theorem

Assume a measurement model $y_k = |\langle x, f_k \rangle + \mu_k|^p + \nu_k$ with $\xi = [\text{real}(x) \ \text{imag}(x)]^T \in \mathring{\Omega}_{z_0}$. Then the covariance of any unbiased estimator $\omega : \mathbb{R}^m \rightarrow \mathbb{C}^n$ is bounded below by

$$\text{Cov}[\omega(y); \xi] \geq (\Pi_{z_0} \mathbb{I}(\xi) \Pi_{z_0})^\dagger.$$

If one chooses the global phase so that $\langle \omega(y), x \rangle \geq 0$ then

$$\text{Cov}[\omega(y); \xi] \geq (\mathbb{I}(\xi))^\dagger.$$

Main Results

Prior Works

Prior literature:

- **2012: B.:** Cramer-Rao lower bound in the real case;
Eldar&Mendelson : map α in the real case

$$\|\alpha(x) - \alpha(y)\| \geq C\|x - y\|\|x + y\|.$$

- **2013: Bandeira, Cahill, Mixon, Nelson:** improved the estimate of C .
B.: β bi-Lipschitz in real and complex case.
- **2014: B.&Yang:** Find the exact Lipschitz constant for α in the real case - the constants A_0, B_0 ; **B.&Z.:** constructed a Lipschitz left inverse for β .
- **2015: B.&Z.:** Proved α is bi-Lipschitz in the complex case; constructed a Lipschitz left inverse. **B.:** lower Lipschitz constant A_0 connected to CRLB of a non-AWGN model.

Main Results

Key relationship between deterministic and stochastic bounds

The central object: $\mathcal{R}(\xi) = \sum_{k=1}^m \Phi_k \xi \xi^T \Phi_k$.

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The lower Lipschitz bound for β map is:

$$a_0 = \min_{\|\xi\|=1} \lambda_{2n-1}(\mathcal{R}(\xi)).$$

The Fisher information matrix for the AWGN model:

$$\mathbb{I}^{AWGN, cplx}(x) = \frac{4}{\sigma^2} \mathcal{R}(\xi).$$

Main Results

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The Fisher information matrix for the AWGN model:

$$\mathbb{I}^{AWGN, cplx}(x) = \frac{4}{\sigma^2} \mathcal{R}(\xi).$$

Best inversion scheme ψ that is lossless in the absence of noise achieves:

$$d_1(\psi(c), \psi(d))^2 \leq \frac{68}{a_0} \|c - d\|_2^2.$$

An efficient estimator (i.e. unbiased that achieves CRLB) ω^0 achieves:

$$\mathbb{E} \left[\|\omega^0(y) - x\|_2^2; x \right] \leq \frac{(2n-1)\sigma^2}{4a_0\|x\|^2} = \frac{2n-1}{4a_0 SNR}.$$

Proofs

Overview

Deterministic bounds: The proofs involve several steps (details in [BZ15]).

- ① Part 1: Injectivity $\longrightarrow$ bi-Lipschitz: Upper bounds are not too hard; lower bounds: relatively easy for β (the "square" map), but relatively hard for α .
- ② Part 2: Left inverse construction is done in three steps:
 - ① The left inverse is first extended to $\mathbb{R}^m$ into $Sym(H)$ using Kirszbraun's theorem;
 - ② Then we show that $\mathcal{S}^{1,0}(H)$ is a Lipschitz retract in $Sym(H)$;
 - ③ The proof is concluded by composing the two maps.

The stochastic bounds: Direct computations and a bit of luck! [Bal15]

Proofs

Part 1a: Bi-Lipschitzianity of α

$$\alpha : \hat{H} \rightarrow \mathbb{R}^m, \quad \alpha(x) = (|\langle x, f_k \rangle|)_{1 \leq k \leq m}$$

The homogeneity of α shows that

$$L(x, y) = \frac{\|\alpha(x) - \alpha(y)\|}{D_p(x, y)}$$

is homogeneous of degree 0: $L(tx, ty) = L(x, y)$, for every $t > 0$.

This reduces the problem to the unit ball: $1 = \|x\| \geq \|y\|$.

The upper bound was computed in [BCMN13]:

$$\sup_{x \neq y} \frac{\|\alpha(x) - \alpha(y)\|^2}{D_2(x, y)^2} = B \quad (\text{upper frame bound}).$$

Proofs

Part 1a: Bi-Lipschitzianity of α - cont'd

A compactness argument shows the lower bound is positive if and only if the local lower bound is positive:

$$\inf_{\|z\|=1} \lim_{r \rightarrow 0} \inf_{\substack{x, y \in \hat{H} \\ D_2(x, z) < r \\ D_2(y, z) < r}} \frac{\|\alpha(x) - \alpha(y)\|_2^2}{D_2(x, y)^2} > 0.$$

This bound is computed explicitly and shown positive: Computations involve the realification framework and other delicate nonlinear expansions.

Proofs

Part 1b: Bi-Lipschitzianity of β

Key Remark (B.Bodmann,Casazza,Edidin - 2007): The nonlinear map β is the restriction of the linear map

$$\mathbb{A} : \text{Sym}(H) \rightarrow \mathbb{R}^m \quad , \quad \mathbb{A}(T) = (\langle Tf_k, f_k \rangle)_{1 \leq k \leq m}$$

Specifically: $\beta(x) = \mathbb{A}(xx^*) = (|\langle x, f_k \rangle|^2)_{1 \leq k \leq m}$.

$$\begin{aligned} \|\beta(x) - \beta(y)\| &= \|\mathbb{A}(xx^*) - \mathbb{A}(yy^*)\| = \|\mathbb{A}(xx^* - yy^*)\| \\ &= \|xx^* - yy^*\| \|\mathbb{A}\left(\frac{xx^* - yy^*}{\|xx^* - yy^*\|}\right)\| \end{aligned}$$

$$a_0 = \min_{T \in \mathcal{S}^{1,1}, \|T\|_1=1} \|\mathbb{A}(T)\| > 0 \quad , \quad b_0 = \max_{T \in \mathcal{S}^{1,1}, \|T\|_1=1} \|\mathbb{A}(T)\|$$

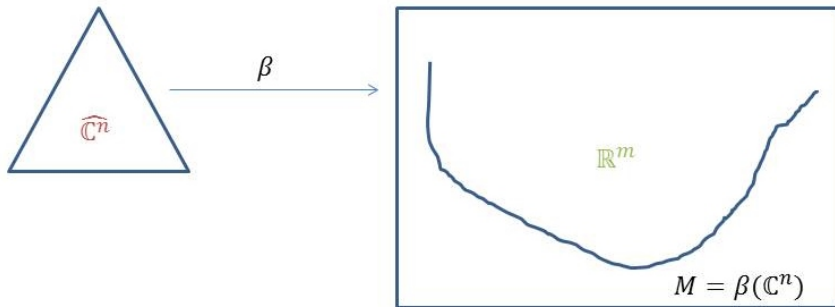
Proofs

Part 2a: Extension of the inverse for α

We know $\alpha : (\hat{H}, D_2) \rightarrow (\mathbb{R}^m, \|\cdot\|_2)$ is bi-Lipschitz:

$$A_0 D_1(x, y)^2 \leq \|\alpha(x) - \alpha(y)\|^2 \leq b_0 D_2(x, y)^2$$

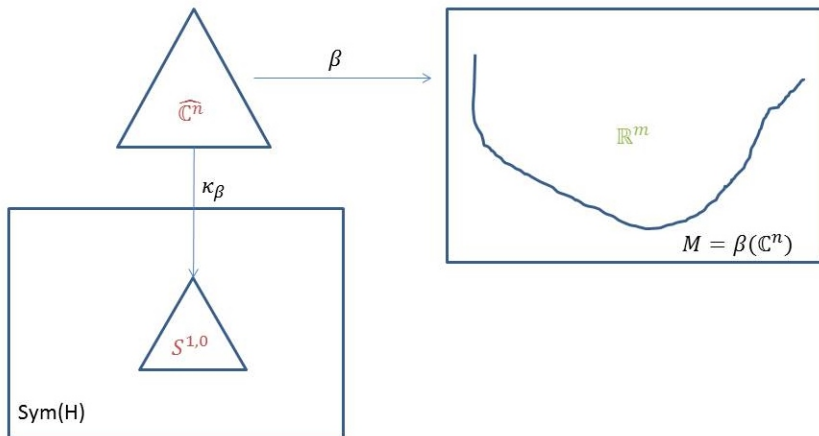
Let $M = \alpha(\hat{H}) \subset \mathbb{R}^m$.



Proofs

Part 2a: Extension of the inverse for α

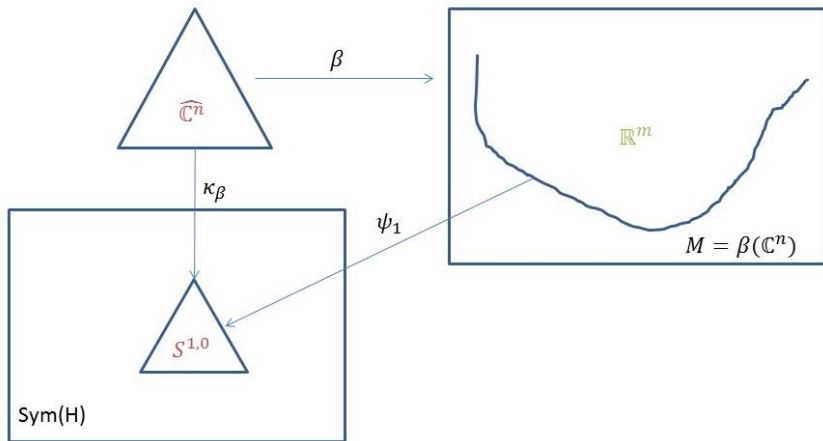
First identify $\hat{H}$ with $\mathcal{S}^{1,0}(H)$.



Proofs

Part 2a: Extension of the inverse for α

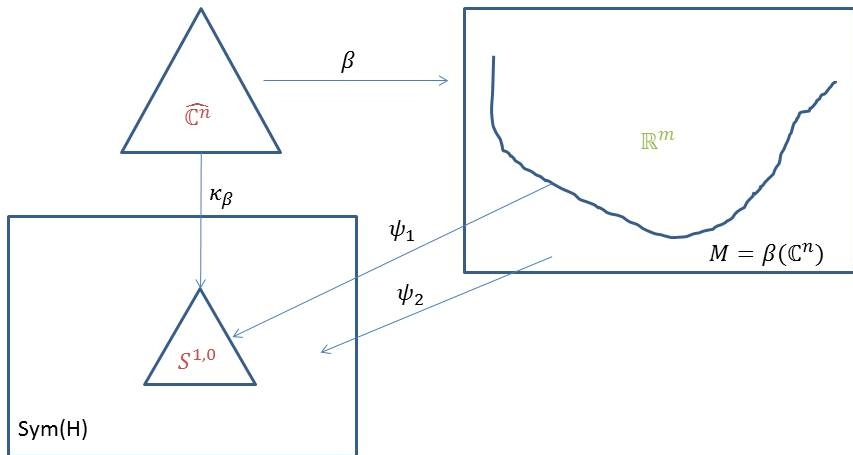
Then construct the local left inverse $\omega_1 : M \rightarrow \hat{H}$ with $Lip(\omega_1) = \frac{1}{\sqrt{A_0}}$.



Proofs

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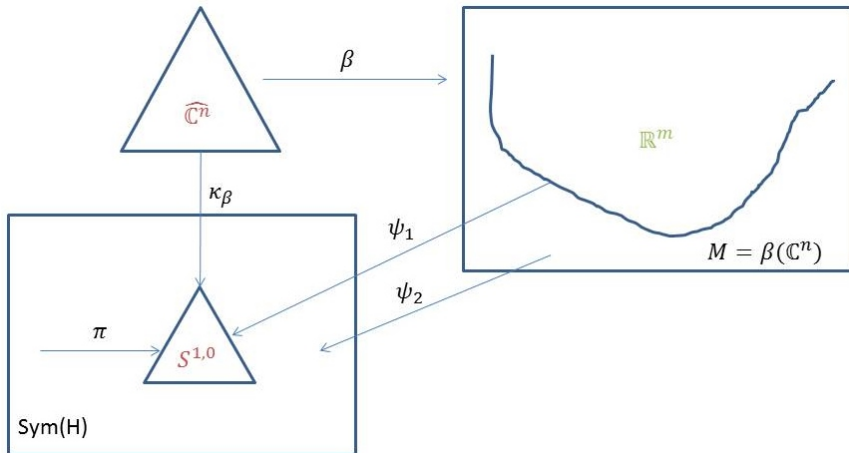
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Proofs

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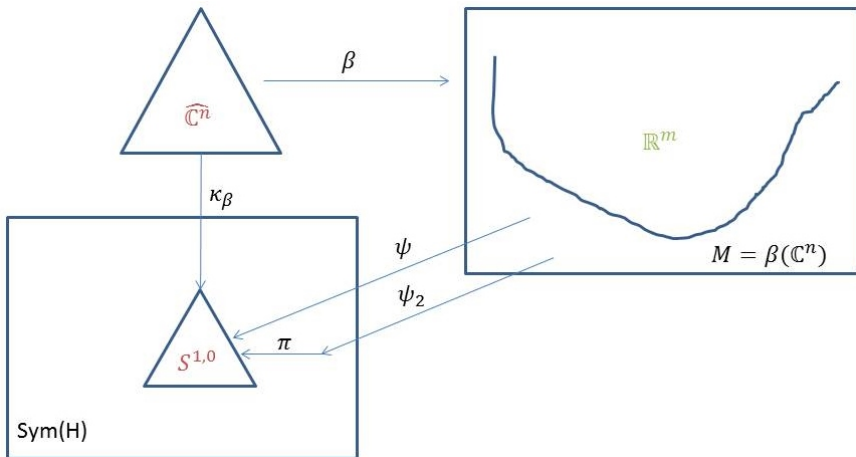
Construct a Lipschitz "projection" $\pi : \text{Sym}(H) \rightarrow \mathcal{S}^{1,0}(H)$.



Proofs

Part 2a: Extension of the inverse for α

Compose the two maps to get $\omega : \mathbb{R}^m \rightarrow \mathcal{S}^{1,0}$, $\omega = \pi \circ \omega_2$.



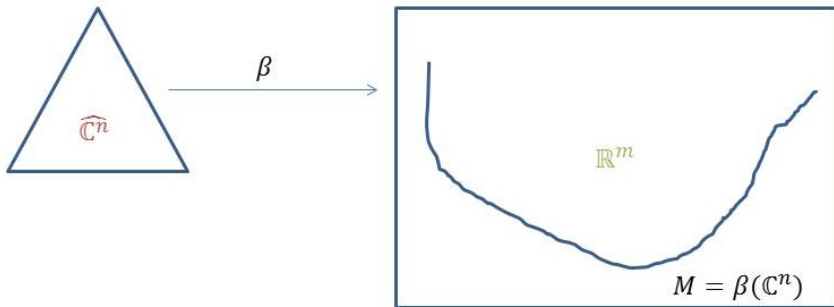
Proofs

Part 2b: Extension of the inverse for β

We know $\beta : (\hat{H}, d_1) \rightarrow (\mathbb{R}^m, \|\cdot\|_2)$ is bi-Lipschitz:

$$a_0 d_1(x, y)^2 \leq \|\beta(x) - \beta(y)\|^2 \leq b_0 d_1(x, y)^2.$$

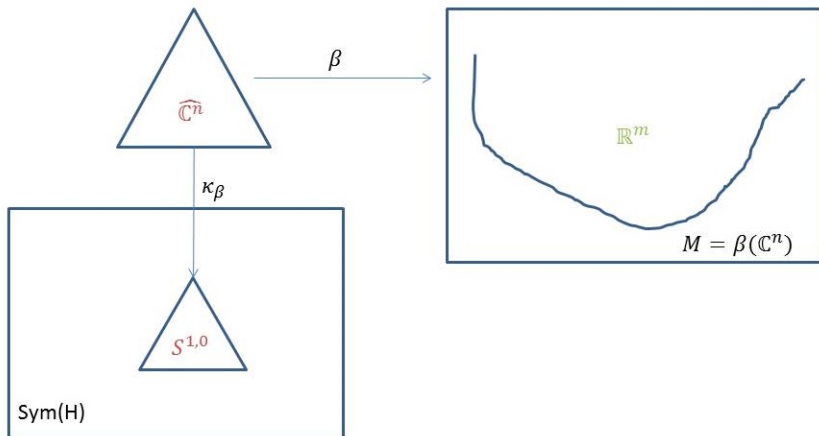
Let $M = \beta(\hat{H}) \subset \mathbb{R}^m$.



Proofs

Part 2b: Extension of the inverse for β

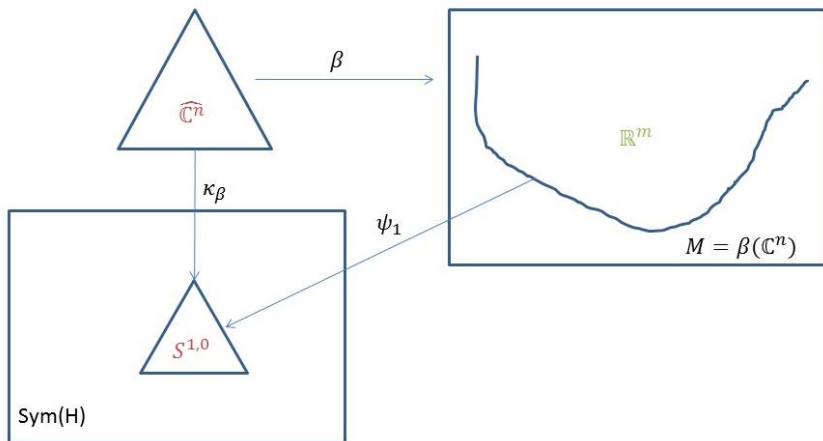
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Proofs

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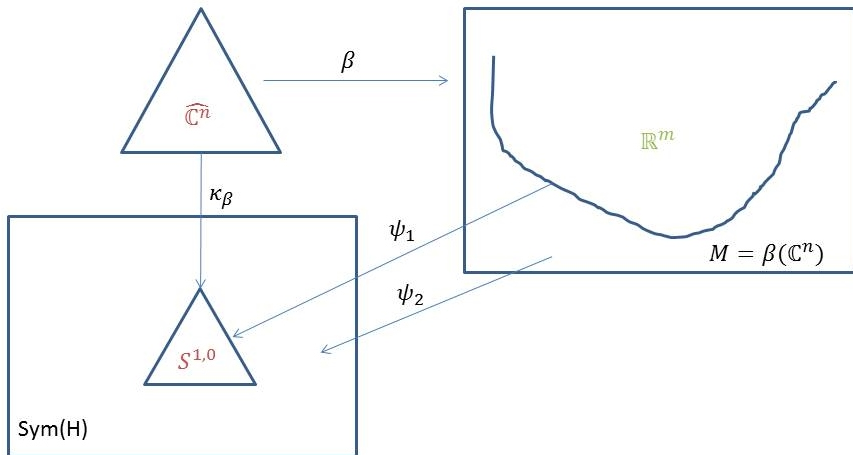
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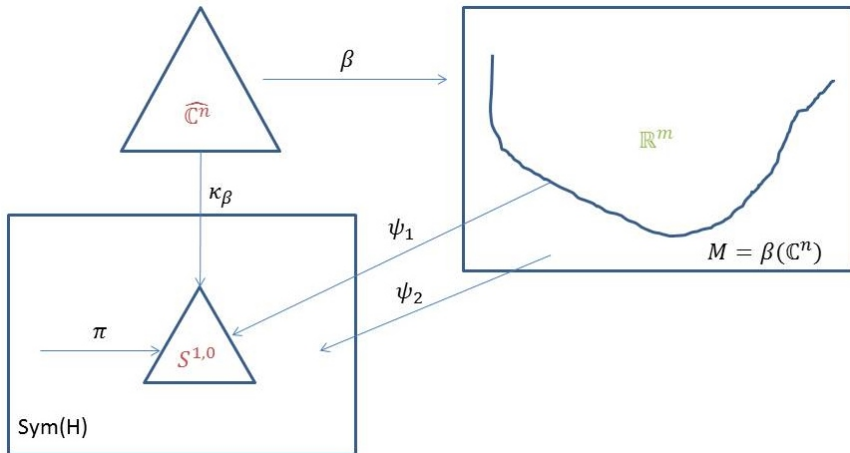
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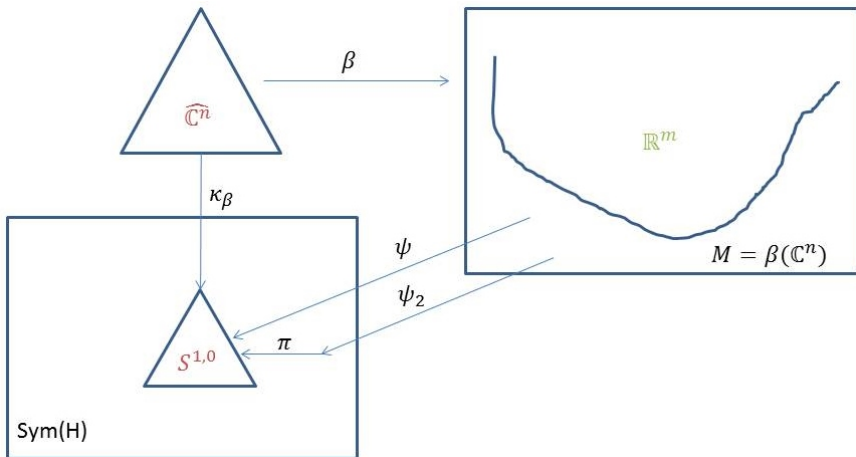
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Proofs

Part 2b: Extension of the inverse for β

Compose the two maps to get $\psi : \mathbb{R}^m \rightarrow \mathcal{S}^{1,0}$, $\psi = \pi \circ \psi_2$.



Proofs

Part 2: $\mathcal{S}^{1,0}(H)$ as Lipschitz retract in $\text{Sym}(H)$

Lemma

Consider the spectral decomposition of the self-adjoint operator A in $\text{Sym}(H)$, $A = \sum_{k=1}^d \lambda_{m(k)} P_k$. Then the map

$$\pi : \text{Sym}(H) \rightarrow \mathcal{S}^{1,0}(H) \quad , \quad \pi(A) = (\lambda_1 - \lambda_2) P_1$$

satisfies the following two properties:

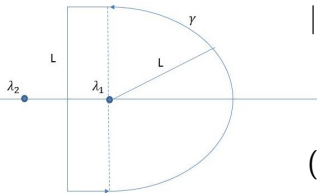
- ① for $1 \leq p \leq \infty$, it is Lipschitz continuous from $(\text{Sym}(H), \|\cdot\|_p)$ to $(\mathcal{S}^{1,0}(H), \|\cdot\|_p)$ with Lipschitz constant less than or equal to $3 + 2^{1+\frac{1}{p}}$;
- ② $\pi(A) = A$ for all $A \in \mathcal{S}^{1,0}(H)$.

Proof uses Weyl's inequality and spectral formula on a complex integration contour by Zwald & Blanchard (2006).

Assume simple top eigenvalues (otherwise the bound is immediate):

$\pi(A) = (\lambda_1 - \lambda_2)P_1$, $\pi(B) = (\mu_1 - \mu_2)Q_1$. Then:

$$\begin{aligned} \|\pi(A) - \pi(B)\|_p &\leq (\lambda_1 - \lambda_2)\|P_1 - Q_1\|_p + |\lambda_1 - \mu_1| + |\lambda_2 - \mu_2| \\ &\leq (\lambda_1 - \lambda_2)\|P_1 - Q_1\|_p + 2\|A - B\|_p. \end{aligned}$$



$$\|P_1 - Q_1\|_p \leq \frac{1}{2\pi} \int_I \|(R_A - R_B)(\gamma(t))\|_p |\gamma'(t)| dt$$

$$R_A(z) = (A - zI)^{-1}, \quad R_B(z) = (B - zI)^{-1}.$$

$$(R_A - R_B)(z) = \sum_{n \geq 1} (-1)^n (R_A(z)(B - A))^n R_A(z).$$

$$\begin{aligned} \|(R_A - R_B)(\gamma(t))\|_p &\leq \sum_{n \geq 1} \|R_A(\gamma(t))\|_\infty^{n+1} \|A - B\|_p^n \\ &= \frac{\|R_A(\gamma(t))\|_\infty^2 \|A - B\|_p}{1 - \|R_A(\gamma(t))\|_\infty \|A - B\|_p} < \frac{\|A - B\|_p}{\text{dist}^2(\gamma(t), \text{Spec}(A))}. \end{aligned}$$

Proofs

Part 1: Bi-Lipschitzianity of α -cont'd

The analysis requires a deeper understanding of local behavior.

- ① The *global lower and upper Lipschitz bounds*:

$$A_0 = \inf_{x,y \in \hat{H}} \frac{\|\alpha(x) - \alpha(y)\|_2^2}{D_2(x,y)^2}, \quad B_0 = \sup_{x,y \in \hat{H}} \frac{\|\alpha(x) - \alpha(y)\|_2^2}{D_2(x,y)^2}$$

- ② The *type I local lower and upper Lipschitz bounds* at $z \in \hat{H}$:

$$A(z) = \lim_{r \rightarrow 0} \inf_{\substack{x,y \in \hat{H} \\ D_2(x,z) < r \\ D_2(y,z) < r}} \frac{\|\alpha(x) - \alpha(y)\|_2^2}{D_2(x,y)^2}, \quad B(z) = \lim_{r \rightarrow 0} \sup_{\substack{x,y \in \hat{H} \\ D_2(x,z) < r \\ D_2(y,z) < r}} \frac{\|\alpha(x) - \alpha(y)\|_2^2}{D_2(x,y)^2}$$

- ③ The *type II local lower and upper Lipschitz bounds* at $z \in \hat{H}$:

$$\tilde{A}(z) = \lim_{r \rightarrow 0} \inf_{\substack{x \in \hat{H} \\ D_2(x,z) < r}} \frac{\|\alpha(x) - \alpha(z)\|_2^2}{D_2(x,z)^2}, \quad \tilde{B}(z) = \lim_{r \rightarrow 0} \sup_{\substack{x \in \hat{H} \\ D_2(x,z) < r}} \frac{\|\alpha(x) - \alpha(z)\|_2^2}{D_2(x,z)^2}$$

Proofs

Part 1: Bi-Lipschitzianity of α -cont'd

We need to analyze the real structure of $\hat{H}$.

Let $\varphi_1, \dots, \varphi_m, \zeta \in \mathbb{R}^{2n}$, $\Phi_1, \dots, \Phi_m \in \text{Sym}(\mathbb{R}^{2n})$, $J \in \mathbb{R}^{2n \times 2n}$ defined by:

$$\Phi_k = \varphi_k \varphi_k^T + J \varphi_k \varphi_k^T J^T, \varphi_k = \begin{bmatrix} \text{real}(f_k) \\ \text{imag}(f_k) \end{bmatrix}, J = \begin{bmatrix} 0 & -I_n \\ I_n & 0 \end{bmatrix}, \zeta = \begin{bmatrix} \text{real}(z) \\ \text{imag}(z) \end{bmatrix}$$

Key relations: $\langle z, f_k \rangle = \langle \zeta, \varphi_k \rangle + i \langle \zeta, J \varphi_k \rangle$, $|\langle z, f_k \rangle| = \sqrt{\langle \Phi_k \zeta, \zeta \rangle}$.

Consider the following objects:

$$\mathcal{R} : \mathbb{R}^{2n} \rightarrow \text{Sym}(\mathbb{R}^{2n}) \quad , \quad \mathcal{R}(\xi) = \sum_{k=1}^m \Phi_k \xi \xi^T \Phi_k \quad , \quad \xi \in \mathbb{R}^{2n}$$

$$\mathcal{S} : \mathbb{R}^{2n} \rightarrow \text{Sym}(\mathbb{R}^{2n}) \quad , \quad \mathcal{S}(\xi) = \sum_{k: \Phi_k \xi \neq 0} \frac{1}{\langle \Phi_k \xi, \xi \rangle} \Phi_k \xi \xi^T \Phi_k \quad , \quad \xi \in \mathbb{R}^{2n}$$

Proofs

Lipschitz bounds for α

Theorem (BZ15)

Assume $\mathcal{F}$ is phase retrievable for $H = \mathbb{C}^n$ and A, B are its optimal frame bounds. Then:

- ① For every $0 \neq z \in \mathbb{C}^n$, $A(z) = \lambda_{2n-1}(\mathcal{S}(\zeta))$ (the next to the smallest eigenvalue);
- ② $A_0 = A(0) > 0$;
- ③ For every $z \in \mathbb{C}^n$, $\tilde{A}(z) = \lambda_{2n-1}(\mathcal{S}(\zeta) + \sum_{k:\langle z, f_k \rangle = 0} \Phi_k)$ (the next to the smallest eigenvalue);
- ④ $\tilde{A}(0) = A$, the optimal lower frame bound;
- ⑤ For every $z \in \mathbb{C}^n$, $B(z) = \tilde{B}(z) = \lambda_1(\mathcal{S}(\zeta) + \sum_{k:\langle z, f_k \rangle = 0} \Phi_k)$ (the largest eigenvalue);
- ⑥ $B_0 = B(0) = \tilde{B}(0) = B$, the optimal upper frame bound;

Proofs

Lipschitz bounds for β

Theorem (cont'd)

- 7 For every $0 \neq z \in \mathbb{C}^n$, $a(z) = \tilde{a}(z) = \lambda_{2n-1}(\mathcal{R}(\zeta))/\|z\|^2$ (the next to the smallest eigenvalue);
- 8 For every $0 \neq z \in \mathbb{C}^n$, $b(z) = \tilde{b}(z) = \lambda_1(\mathcal{R}(\zeta))/\|z\|^2$ (the largest eigenvalue);
- 9 $a_0 = \min_{\|\xi\|=1} \lambda_{2n-1}(\mathcal{R}(\xi))$ is also the largest constant to that $\mathcal{R}(\xi) \geq a_0(\|\xi\|^2 I - J\xi\xi^T J^T)$;
- 10 $b(0) = \tilde{b}(0) = b_0 = \max_{\|\xi\|=1} \lambda_1(\mathcal{R}(\xi))$ is also the 4th power of the frame analysis operator norm $T : (\mathbb{C}^n, \|\cdot\|_2) \rightarrow (\mathbb{R}^m, \|\cdot\|_4)$:
 $b_0 = \|T\|_{B(l^2, l^4)}^4 = \max_{\|x\|_2=1} \sum_{k=1}^m |\langle x, f_k \rangle|^4$;
- 11 $\tilde{a}(0)$ is given by $\tilde{a}(0) = \min_{\|z\|=1} \sum_{k=1}^m |\langle z, f_k \rangle|^4$.

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