

Orthogonality to matrix subspaces

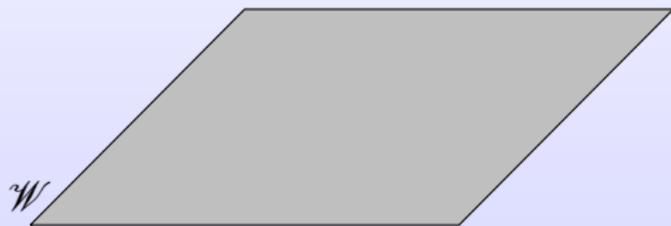
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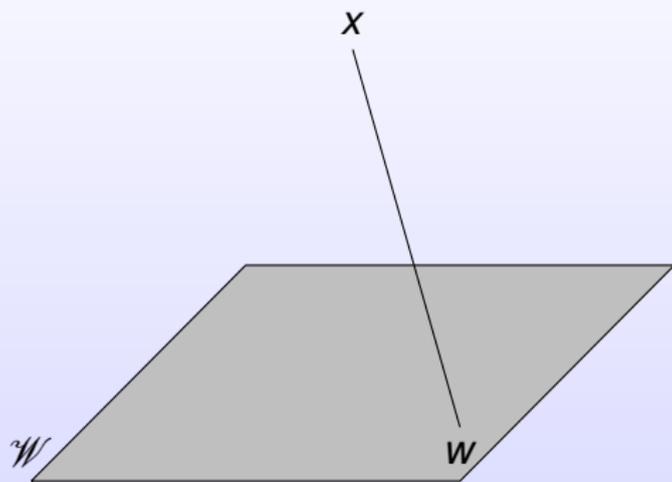
Motivation

Let \mathcal{W} be a subspace of \mathbb{R}^n .



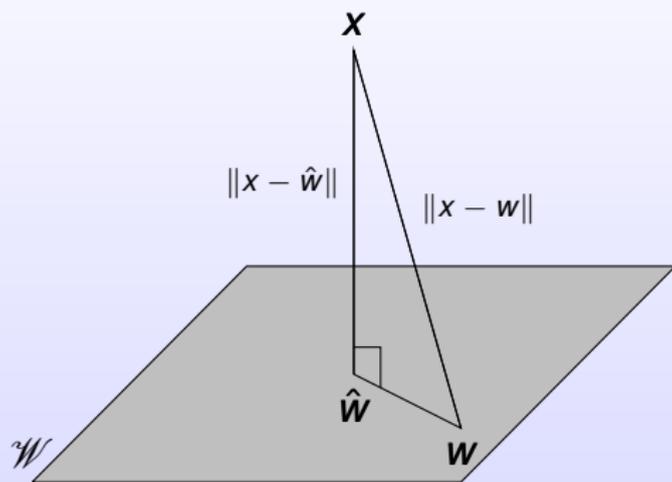
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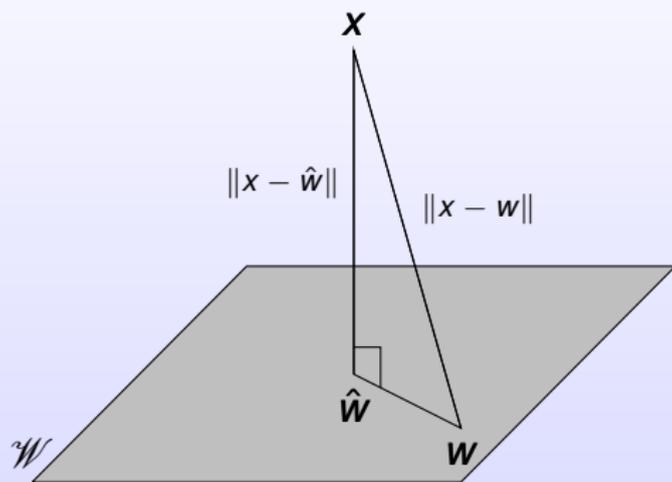


The orthogonal projection of x onto \mathcal{W} , namely \hat{w} is the closest point in \mathcal{W} to x .

$$\|x - \hat{w}\| < \|x - w\| \text{ for all } w \in \mathcal{W}, w \neq \hat{w}$$

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$$\|x - \hat{w}\| < \|x - w\| \text{ for all } w \in \mathcal{W}, w \neq \hat{w}$$

\hat{w} is called the **best approximation** from \mathcal{W} to x .

Motivation

Let $\mathbb{M}(n)$ be the space of $n \times n$ complex matrices. Let \mathcal{W} be a subspace of $\mathbb{M}(n)$. Let $A \notin \mathcal{W}$. Consider the problem of finding a best approximation from \mathcal{W} to A .

That is, find an element \hat{W} such that

$$\min_{W \in \mathcal{W}} \|A - W\| = \|A - \hat{W}\|.$$

A specific question: When is zero a best approximation from \mathcal{W} to A ? That is, when do we have

$$\min_{W \in \mathcal{W}} \|A - W\| = \|A\|?$$

Suppose \mathcal{W} is the subspace spanned by a single element B . Then the above problem reduces to

$$\min_{\lambda \in \mathbb{C}} \|A - \lambda B\| = \|A\|.$$

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Definition

A is said to be **Birkhoff-James orthogonal** to B if

$$\min_{\lambda \in \mathbb{C}} \|A - \lambda B\| = \|A\|,$$

that is,

$$\|A + \lambda B\| \geq \|A\| \text{ for all } \lambda \in \mathbb{C}.$$

In general, if \mathcal{X} is a complex normed linear space, then an element x is said to be Birkhoff-James orthogonal to another element y if

$$\|x + \lambda y\| \geq \|x\| \text{ for all } \lambda \in \mathbb{C}.$$

If \mathcal{X} is a pre-Hilbert space with inner product $\langle \cdot, \cdot \rangle$, then this definition is equivalent to $\langle x, y \rangle = 0$.

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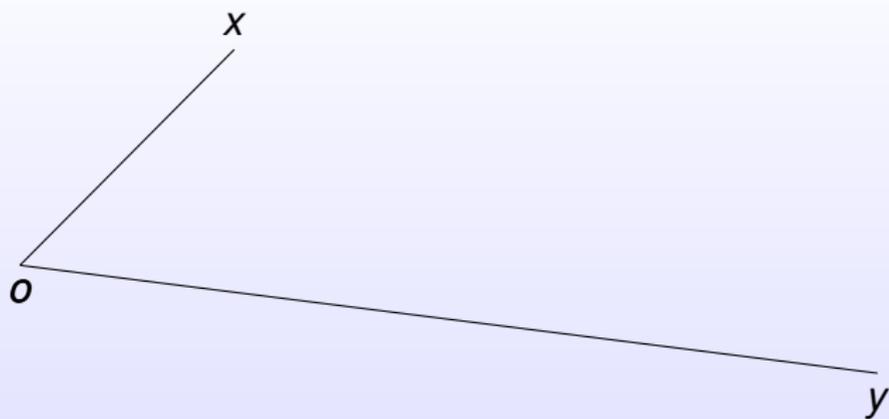
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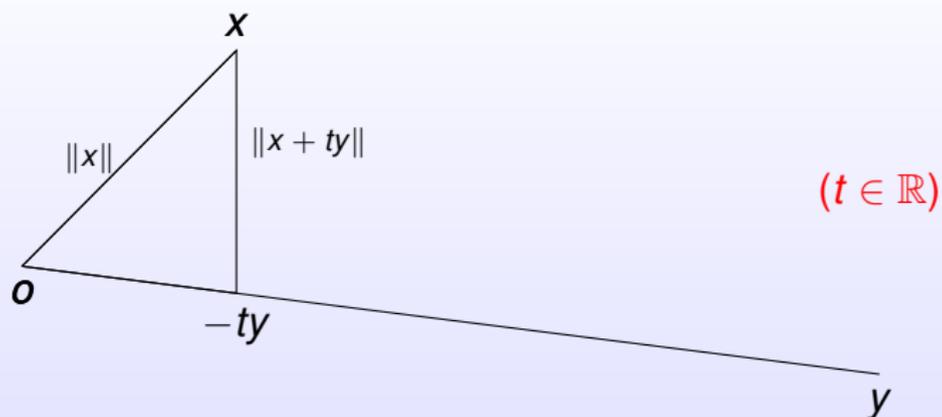
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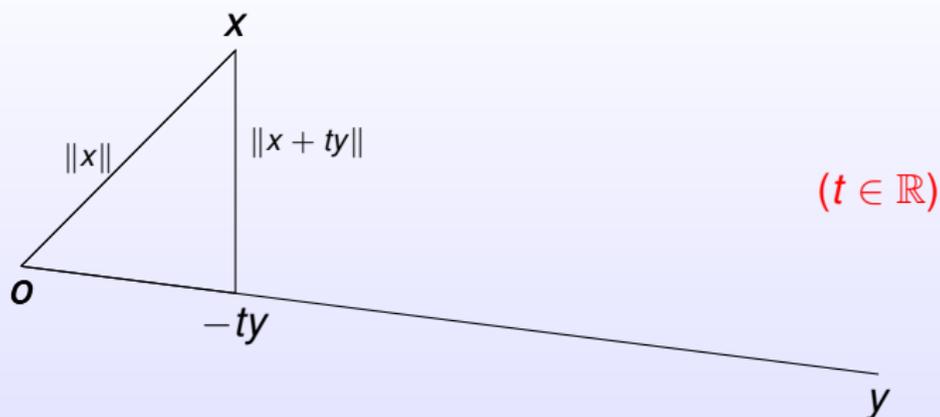
Geometric interpretation



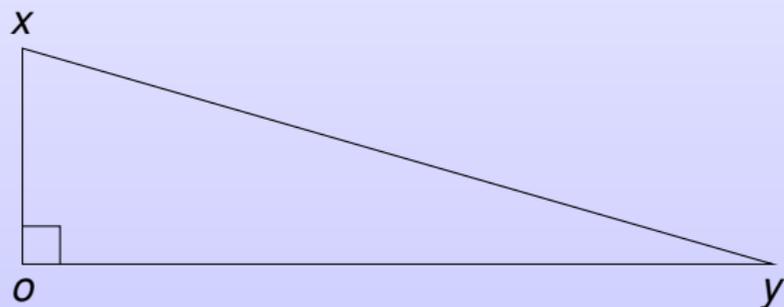
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Geometric interpretation



$\|x + ty\| \geq \|x\|$ for all $t \in \mathbb{R}$ if and only if x and y are orthogonal.



\mathcal{X} real Banach space
 $x, y \in \mathcal{X}$

Observation

Either $\|x + ty\| \geq \|x\|$ for all $t \geq 0$ or $\|x + ty\| \geq \|x\|$ for all $t \leq 0$.

x is Birkhoff-James orthogonal to y if both of them are satisfied.

Properties

- This orthogonality is clearly homogeneous: x orthogonal to $y \Rightarrow \lambda x$ orthogonal to μy for all scalars λ, μ .
- Not symmetric: x orthogonal to $y \not\Rightarrow y$ orthogonal to x .
- Not additive: x orthogonal to $y, z \not\Rightarrow x$ orthogonal to $y + z$.

$$\|x + ty\| \geq \|x\| \text{ for all } t \in \mathbb{R}$$

- Let $f(t) = \|x + ty\|$ mapping \mathbb{R} into \mathbb{R}_+ .
- To say that $\|x + ty\| \geq \|x\|$ for all $t \in \mathbb{R}$ is to say that f attains its minimum at the point 0.
- A calculus problem?
- If f were differentiable at x , then a necessary and sufficient condition for this would have been that the derivative $Df(0) = 0$.
- But the norm function may not be differentiable at x .
- However, f is a convex function, that is,
$$f(\alpha x + (1-\alpha)y) \leq \alpha f(x) + (1-\alpha)f(y) \text{ for all } x, y \in \mathcal{X}, 0 \leq \alpha \leq 1.$$
- The tools of convex analysis are available.

Orthogonality in matrices

$\mathbb{M}(n)$: the space of $n \times n$ complex matrices

$$\langle A, B \rangle = \text{tr}(A^* B)$$

$\|\cdot\|$ is the operator norm, $\|A\| = \sup_{\|x\|=1} \|Ax\|$.

Theorem (Bhatia, Šemrl; 1999)

Let $A, B \in \mathbb{M}(n)$. Then $A \perp_{BJ} B$ if and only if there exists $x : \|x\| = 1$, $\|Ax\| = \|A\|$ and $\langle Ax, Bx \rangle = 0$.

Importance: It connects the more complicated Birkhoff-James orthogonality in the space $\mathbb{M}(n)$ to the standard orthogonality in the space \mathbb{C}^n .

Bhatia-Šemrl Theorem

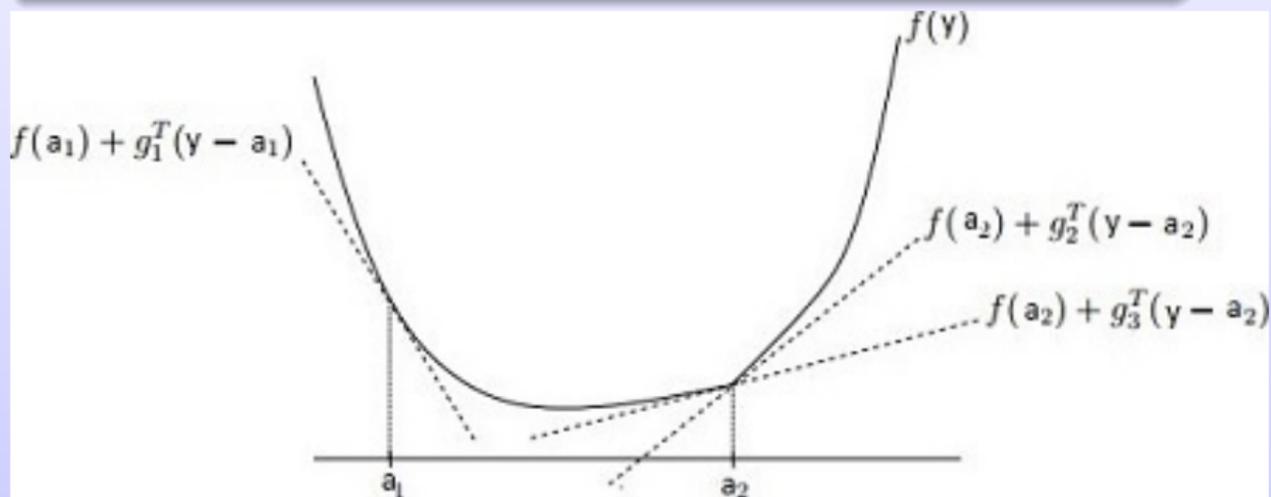
- Let $f : t \rightarrow \|A + tB\|$, $A \geq 0$.
- $A \perp_{BJ} B$ if and only if f attains its minimum at 0.
- f is a convex function but may not be differentiable.

Subdifferential

Definition 1

Let $f : \mathcal{X} \rightarrow \mathbb{R}$ be a convex function. The *subdifferential* of f at a point $a \in \mathcal{X}$, denoted by $\partial f(a)$, is the set of continuous linear functionals $\varphi \in \mathcal{X}^*$ such that

$$f(y) - f(a) \geq \operatorname{Re} \varphi(y - a) \quad \text{for all } y \in \mathcal{X}.$$

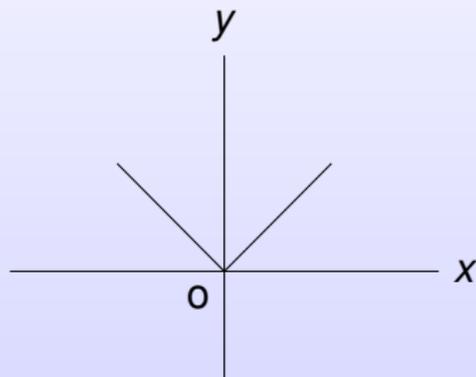


Examples

Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be defined as

$$f(x) = |x|.$$

This function is differentiable at all $a \neq 0$ and $D f(a) = \text{sign}(a)$.
At zero, it is not differentiable.

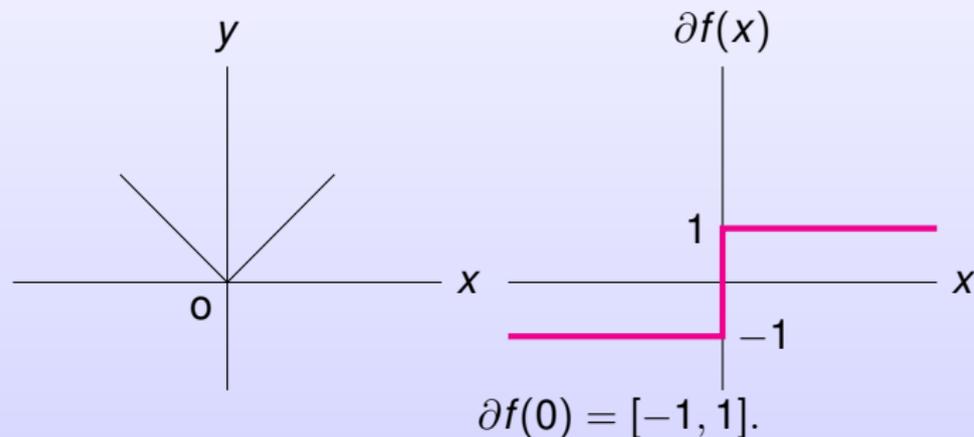


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Note that for $v \in \mathbb{R}$,

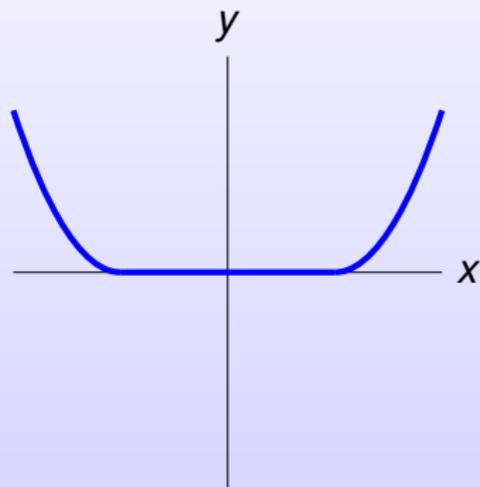
$$f(y) = |y| \geq f(0) + v \cdot y = v \cdot y$$

holds for all $y \in \mathbb{R}$ if and only if $|v| \leq 1$.

Examples

Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be the map defined as

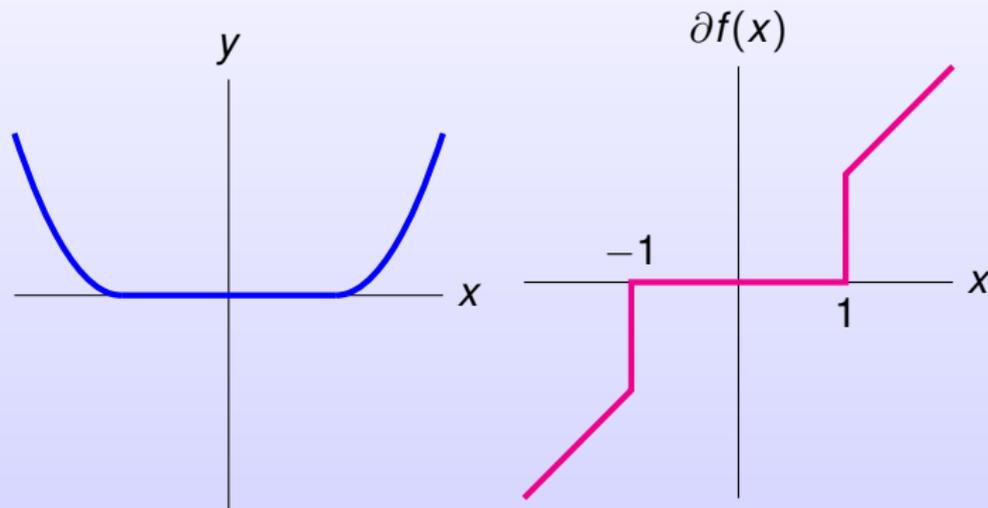
$$f(x) = \max \left\{ 0, \frac{x^2 - 1}{2} \right\}.$$



Examples

Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be the map defined as

$$f(x) = \max \left\{ 0, \frac{x^2 - 1}{2} \right\}.$$



Then f is differentiable everywhere except at $x = -1, 1$. We have

$$\partial f(1) = [0, 1] \text{ and } \partial f(-1) = [-1, 0].$$

Examples

Let $f : \mathcal{X} \rightarrow \mathbb{R}$ be defined as

$$f(a) = \|a\|.$$

Then for $a \neq 0$,

$$\partial f(a) = \{\varphi \in \mathcal{X}^* : \operatorname{Re} \varphi(a) = \|a\|, \|\varphi\| \leq 1\},$$

and

$$\partial f(0) = \{\varphi \in \mathcal{X}^* : \|\varphi\| \leq 1\}.$$

In particular, when $\mathcal{X} = \mathbb{M}(n)$, we get that for $A \neq 0$,

$$\partial f(A) = \{G \in \mathbb{M}(n) : \operatorname{Re} \operatorname{tr} G^* A = \|A\|, \|G\|_* \leq 1\}.$$

($\|\cdot\|_*$ is the dual norm of $\|\cdot\|$)

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be defined as

$$f(\mathbf{a}) = \|\mathbf{a}\|_\infty = \max\{|\mathbf{a}_1|, \dots, |\mathbf{a}_n|\}.$$

Then for $\mathbf{a} \neq \mathbf{0}$,

$$\partial f(\mathbf{a}) = \text{conv}\{\pm \mathbf{e}_i : |\mathbf{a}_i| = \|\mathbf{a}\|_\infty\}.$$

$$f(y) - f(a) \geq \operatorname{Re} \varphi(y - a) \quad \text{for all } y \in \mathcal{X}.$$

Proposition

A function $f : \mathcal{X} \rightarrow \mathbb{R}$ attains its minimum value at $a \in \mathcal{X}$ if and only if $0 \in \partial f(a)$.

In our case,

$$f(t) = \|A + tB\|$$

and f attains minimum at 0. So

$$0 \in \partial f(0).$$

Precomposition with an affine map

Let \mathcal{X}, \mathcal{Y} be any two Banach spaces. Let $g : \mathcal{Y} \rightarrow \mathbb{R}$ be a continuous convex function. Let $S : \mathcal{X} \rightarrow \mathcal{Y}$ be a linear map and let $L : \mathcal{X} \rightarrow \mathcal{Y}$ be the continuous affine map defined by $L(x) = S(x) + y_0$, for some $y_0 \in \mathcal{Y}$. Then

$$\partial(g \circ L)(a) = S^* \partial g(L(a)) \text{ for all } a \in \mathcal{X}.$$

In our case, let $S : t \mapsto tB$, $L : t \mapsto A + tB$ and

$$g : X \mapsto \|X\|.$$

And $f(t) = \|A + tB\|$ is the composition of g and L . So

$$\partial f(0) = S^* \partial \|A\|,$$

where $S^*(T) = \operatorname{Re} \operatorname{tr} B^* T$.

Watson, 1992

For any $A \in \mathbb{M}(n)$,

$$\partial\|A\| = \text{conv}\{uv^* : \|u\| = \|v\| = 1, Av = \|A\|u\}.$$

If $A \geq 0$, then

$$\partial\|A\| = \text{conv}\{uu^* : \|u\| = 1, Au = \|A\|u\}.$$

Bhatia-Šemrl Theorem

- $0 \in \partial f(0) = S^* \partial \|A\|$ if and only if
 $0 \in \text{conv}\{ \text{Re} \langle u, Bu \rangle : \|u\| = 1, Au = \|A\|u \}$.
- By Hausdorff-Toeplitz Theorem,
 $\{ \text{Re} \langle u, Bu \rangle : \|u\| = 1, Au = \|A\|u \}$ is convex.
- $0 \in S^* \partial \|A\|$ if and only if
 $0 \in \{ \text{Re} \langle u, Bu \rangle : \|u\| = 1, Au = \|A\|u \}$.
- There exists $x : \|x\| = 1, Ax = \|A\|x$ and $\text{Re} \langle Ax, Bx \rangle = 0$.

Distance of A from $\mathbb{C}I$:

$$\text{dist}(A, \mathbb{C}I) = \min\{\|A - \lambda I\| : \lambda \in \mathbb{C}\}$$

Variance of A with respect to x :

For $x : \|x\| = 1$,

$$\text{var}_x(A) = \|Ax\|^2 - |\langle x, Ax \rangle|^2.$$

Corollary

Let $A \in \mathbb{M}(n)$. With notations as above, we have

$$\text{dist}(A, \mathbb{C}I)^2 = \max_{\|x\|=1} \text{var}_x(A).$$

Orthogonality to a subspace

\mathcal{W} : subspace of $\mathbb{M}(n)$

Consider the problem of finding \hat{W} such that

$$\min_{W \in \mathcal{W}} \|A - W\| = \|A - \hat{W}\|$$

Examples

1) Let $\mathcal{W} = \{W : \text{tr } W = 0\}$.

Then

$$\min_{W \in \mathcal{W}} \|A - W\| = \frac{|\text{tr}(A)|}{n}.$$

$$\hat{W} = A - \frac{\text{tr}(A)}{n} I.$$

2) Let $\mathcal{W} = \{W : W = W^*\}$.

For any A ,

$$\hat{W} = \frac{1}{2}(A + A^*) = \text{Re } A.$$

Orthogonality to a subspace

When is zero matrix a best approximation to \mathcal{W} ?

A is said to be Birkhoff-James orthogonal to \mathcal{W} ($A \perp_{BJ} \mathcal{W}$) if

$$\|A + W\| \geq \|A\| \text{ for all } W \in \mathcal{W}.$$

\mathcal{W}^\perp : the orthogonal complement of \mathcal{W} , under the usual Hilbert space orthogonality in $\mathbb{M}(n)$ with the inner product $\langle A, B \rangle = \text{tr}(A^* B)$.

Bhatia-Šemrl theorem: $A \perp_{BJ} \mathbb{C}B$ if and only if there exists a positive semidefinite matrix P of rank one such that $\text{tr } P = 1$, $\text{tr } A^* A P = \|A\|^2$ and $AP \in (\mathbb{C}B)^\perp$.

$\mathbb{D}(n; \mathbb{R})$: the space of real diagonal $n \times n$ matrices

A matrix A is said to be minimal if $\|A + D\| \geq \|A\|$ for all $D \in \mathbb{D}(n; \mathbb{R})$, i.e. A is orthogonal to the subspace $\mathbb{D}(n; \mathbb{R})$.

Theorem (Andruchow, Larotonda, Recht, Varela; 2012)

A Hermitian matrix A is minimal if and only if there exists a $P \geq 0$ such that

$$A^2 P = \|A\|^2 P$$

and

all the diagonal elements of AP are zero.

Question: Similar characterizations for other subspaces?

Theorem

Let $A \in \mathbb{M}(n)$ and let \mathcal{W} be a subspace of $\mathbb{M}(n)$. Then $A \perp_{BJ} \mathcal{W}$ if and only if there exists $P \geq 0$, $\text{tr } P = 1$, such that

$$A^*AP = \|A\|^2 P$$

and

$$AP \in \mathcal{W}^\perp.$$

Moreover, we can choose P such that $\text{rank } P \leq m(A)$, where $m(A)$ is the multiplicity of the maximum singular value $\|A\|$ of A .

Orthogonality to a subspace

$m(A)$ is the best possible upper bound on rank P .

Consider $\mathcal{W} = \{X : \text{tr } X = 0\}$.

Then $\{A : A \perp_{BJ} \mathcal{W}\} = \mathcal{W}^\perp = \mathbb{C}I$.

If $A \perp_{BJ} \mathcal{W}$, then it has to be of the form $A = \lambda I$, for some $\lambda \in \mathbb{C}$.

When $A \neq 0$ then $m(A) = n$.

Let P be any density matrix satisfying $AP \in \mathcal{W}^\perp$. Then $AP = \mu I$, for some $\mu \in \mathbb{C}$, $\mu \neq 0$.

If P also satisfies $A^*AP = \|A\|^2 P$, then we get $P = \frac{\mu}{\lambda} I$. Hence rank $P = n = m(A)$.

Orthogonality to a subspace

Observation: In general, the set $\{A : A \perp_{BJ} \mathcal{W}\}$ need not be a subspace.

Consider the subspace $\mathcal{W} = \mathbb{C}I$ of $\mathbb{M}(3)$. Let

$$A_1 = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix} \text{ and } A_2 = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix}.$$

Then $A_1, A_2 \perp_{BJ} \mathcal{W}$.

$$\text{Then } A_1 + A_2 = \begin{bmatrix} 0 & 1 & 1 \\ 1 & 0 & 1 \\ 1 & 1 & 0 \end{bmatrix}, \|A_1 + A_2\| = 2.$$

But $\|A_1 + A_2 - \frac{3}{2}I\| = \frac{3}{2} < \|A_1 + A_2\|$. Hence $A_1 + A_2 \not\perp_{BJ} \mathcal{W}$.

Distance to any subalgebra of $\mathbb{M}(n)$

$\text{dist}(A, \mathcal{W})$: distance of a matrix A from the subspace \mathcal{W}

$$\text{dist}(A, \mathcal{W}) = \min \{ \|A - W\| : W \in \mathcal{W} \}.$$

We have seen that

$$\text{dist}(A, \mathbb{C}I)^2 = \max_{\|x\|=1} \text{var}_x(A).$$

This is equivalent to saying that

$$\text{dist}(A, \mathbb{C}I)^2 = \max \left\{ \text{tr}(A^*AP) - |\text{tr}(AP)|^2 : P \geq 0, \text{tr} P = 1, \text{rank} P = 1 \right\}.$$

Let \mathcal{B} be any C^* subalgebra of $\mathbb{M}(n)$.

Similar distance formula?

(This question has been raised by Rieffel)

Distance to a subalgebra of $\mathbb{M}(n)$

$\mathcal{C}_{\mathcal{B}} : \mathbb{M}(n) \rightarrow \mathcal{B}$ denote the projection of $\mathbb{M}(n)$ onto \mathcal{B} .

Theorem

For any $A \in \mathbb{M}(n)$

$$\text{dist}(A, \mathcal{B})^2 = \max\{\text{tr}(A^*AP - \mathcal{C}_{\mathcal{B}}(AP)^*\mathcal{C}_{\mathcal{B}}(AP)\mathcal{C}_{\mathcal{B}}(P)^{-1}) \\ : P \geq 0, \text{tr } P = 1\},$$

where $\mathcal{C}_{\mathcal{B}}(P)^{-1}$ denotes the Moore-Penrose inverse of $\mathcal{C}_{\mathcal{B}}(P)$.
The maximum on the right hand side can be restricted to $\text{rank } P \leq m(A)$.

The quantity on the right also enjoys the property of being translation invariant.

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THANK YOU