

# C\*環を用いたカーネル法の拡張

## 2. C\*環を用いたカーネル法の理論的側面

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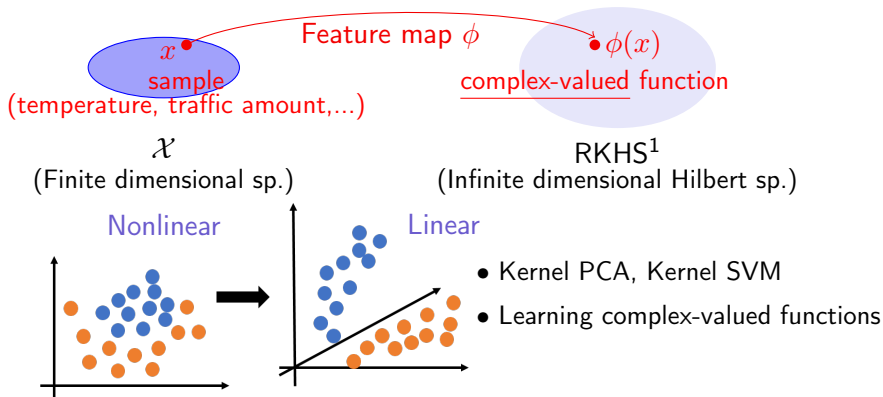
NTT / RIKEN AIP

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- Y. Hashimoto, I. Ishikawa, M. Ikeda, F. Komura, T. Katsura, and Y. Kawahara, JMLR, 22(267):1–56. (updated version : arXiv:2101.11410v2)
- Y. Hashimoto, F. Komura, and M. Ikeda, Matrix and Operator Equations, pp. 1–27.

1. Motivation and Background
2. Reproducing kernel Hilbert  $C^*$ -module (RKHM)
  - 2.1 Definition of RKHM
  - 2.2 Theories for applying RKHM to data analysis
3. Conclusion

# Background: Kernel methods



## Advantages of RKHS

- Nonlinearity in the original space is transformed into a linear one.
- We can compute inner products in RKHS exactly by computers.

<sup>1</sup>Schölkopf and Smola, MIT Press, Cambridge, 2001

# Background: Reproducing kernel Hilbert space (RKHS)

Let  $\mathcal{X}$  be a set. A map  $k : \mathcal{X} \times \mathcal{X} \rightarrow \mathbb{C}$  is called a **positive definite kernel** if it satisfies:

1.  $k(x, y) = \overline{k(y, x)}$  for  $x, y \in \mathcal{X}$  and
2.  $\sum_{t,s=1}^n \overline{c_t} k(x_t, x_s) c_s \geq 0$  for  $n \in \mathbb{N}$ ,  $c_1, \dots, c_n \in \mathbb{C}$ ,  $x_1, \dots, x_n \in \mathcal{X}$ .

$\phi(x) := k(\cdot, x)$  ( $\phi : \mathcal{X} \rightarrow \mathbb{C}^{\mathcal{X}}$ : feature map associated with  $k$ ),

$$\mathcal{H}_{k,0} := \left\{ \sum_{t=1}^n \phi(x_t) c_t \mid n \in \mathbb{N}, c_t \in \mathbb{C}, x_t \in \mathcal{X} \right\}. \quad (1)$$

We can define an **inner product**  $\langle \cdot, \cdot \rangle_k : \mathcal{H}_{k,0} \times \mathcal{H}_{k,0} \rightarrow \mathbb{C}$  as

$$\left\langle \sum_{s=1}^n \phi(x_s) c_s, \sum_{t=1}^l \phi(y_t) d_t \right\rangle_k := \sum_{s=1}^n \sum_{t=1}^l \overline{c_s} k(x_s, y_t) d_t. \quad (2)$$

Reproducing property:  $\langle \phi(x), v \rangle_k = v(x)$  for  $v \in \mathcal{H}_k$  and  $x \in \mathcal{X}$

**RKHS  $\mathcal{H}_k$ : completion of  $\mathcal{H}_{k,0}$**

# Background: Representer theorem in RKHSs

The representer theorem guarantees that solutions of a minimization problem are **represented only with given samples**<sup>2</sup>.

$\mathcal{H}_k$ : RKHS

$$\mathbb{R}_+ := \{a \in \mathbb{R} \mid a \geq 0\}$$

## Theorem 1 Representer theorem in RKHSs

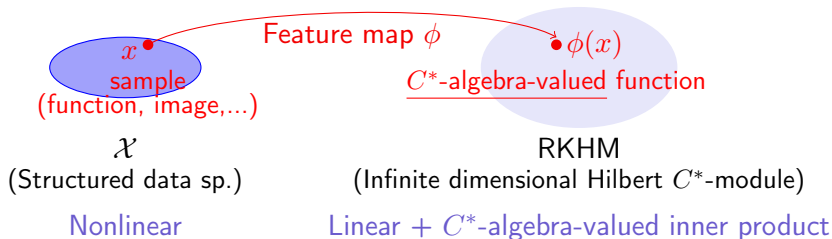
Let  $x_1, \dots, x_n \in \mathcal{X}$  and  $a_1, \dots, a_n \in \mathbb{C}$ . Let  $h : \mathcal{X} \times \mathbb{C}^2 \rightarrow \mathbb{R}_+$  be an error function and  $g : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  satisfy  $g(c) < g(d)$  for  $c < d$ . Then, any  $u \in \mathcal{H}_k$  minimizing  $\sum_{i=1}^n h(x_i, a_i, u(x_i)) + g(\|u\|_k)$  admits a representation of the form  $\sum_{i=1}^n \phi(x_i)c_i$  for some  $c_1, \dots, c_n \in \mathbb{C}$ .

The result can be applied to supervised problems.

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<sup>2</sup>Schölkopf et al., COLT 2001.

# Goal: Generalization of data analysis in RKHS to RKHM



## Advantages of RKHM:

- $C^*$ -algebra-valued inner products extract information of **structures**.

## We constructed a framework of data analysis with RKHM.

- We can reconstruct existing RKHSs by using RKHMs.
- We have shown fundamental properties for data analysis in RKHMs (e.g. representer theorem, kernel mean embedding).

# $C^*$ -algebra and von Neumann-algebra

$C^*$ -algebra : Banach space equipped with a product & an involution  $*$   
+  $C^*$ -property

e.g.

- $C(\mathcal{Z})$  for a compact space  $\mathcal{Z}$   
**Norm** : sup norm, **Product** : pointwise product,  
**Involution** : pointwise complex conjugate
- $\mathcal{K}(\mathcal{H}) = \{\text{compact operators on a Hilbert space } \mathcal{H}\}$   
**Norm** : operator norm, **Product** : composition, **Involution** : adjoint

**Von Neumann-algebra** :  $C^*$ -algebra that is closed in the strong operator topology

e.g.

- $L^\infty(\mathcal{Z})$  for a measure space  $\mathcal{Z}$
- $\mathcal{B}(\mathcal{H}) = \{\text{bounded linear operators on a Hilbert space } \mathcal{H}\}$

# Positivity and order in $C^*$ -algebras

For optimization, we need the notion of “positive” and order.

$\mathcal{A}$ :  $C^*$ -algebra

## Definition 1 Positive

Let  $a \in \mathcal{A}$ . If  $a = b^*b$  for some  $b \in \mathcal{A}$ , then  $a$  is called **positive**. We put  $\mathcal{A}_+ = \{a \in \mathcal{A} \mid a \text{ is positive}\}$ .

We can define a (partial) order  $\leq_{\mathcal{A}}$  in  $\mathcal{A}$  by “ $a \leq_{\mathcal{A}} b$  if and only if  $b - a$  is positive”.

We denote  $a <_{\mathcal{A}} b$  if  $b - a$  is positive and not zero.

We consider supremum, maximum, infimum, and minimum in  $\mathcal{A}$  with respect to the order  $\leq_{\mathcal{A}}$ .

# Hilbert $C^*$ -module

$\mathcal{A}$ :  $C^*$ -algebra

$\mathcal{M}$ : right  $\mathcal{A}$ -module ( $u \in \mathcal{M}, c \in \mathcal{A} \rightarrow uc \in \mathcal{M}$ )

## Definition 2 $\mathcal{A}$ -valued inner product

A map  $\langle \cdot, \cdot \rangle : \mathcal{M} \times \mathcal{M} \rightarrow \mathcal{A}$  is called an  $\mathcal{A}$ -valued inner product if it satisfies the following properties for  $u, v, w \in \mathcal{M}$  and  $c, d \in \mathcal{A}$ :

1.  $\langle u, vc + wd \rangle = \langle u, v \rangle c + \langle u, w \rangle d$ ,
2.  $\langle v, u \rangle = \langle u, v \rangle^*$ ,
3.  $\langle u, u \rangle \geq 0$  (positive) and if  $\langle u, u \rangle = 0$  then  $u = 0$ .

$\rightarrow \mathcal{A}$ -valued absolute value  $|u| := \langle u, u \rangle^{1/2} \rightarrow$  Norm  $\|u\| := \|\langle u, u \rangle\|_{\mathcal{A}}^{1/2}$

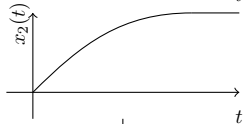
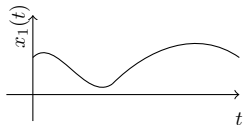
**Hilbert  $C^*$ -module  $\mathcal{M}$ <sup>3</sup>**: complete  $\mathcal{A}$ -module equipped with an  $\mathcal{A}$ -valued inner-product

<sup>3</sup>Lance, Cambridge University Press, 1995.

# Advantages of RKHM (functional data)

## Algorithms in RKHS

$x_1, x_2$  : Functional data  
 $x_1, x_2 \in \mathcal{H}$

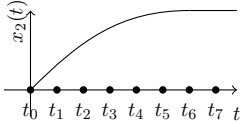
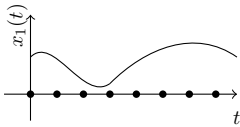


Compute the  
inner product

$$\langle x_1, x_2 \rangle_{\mathcal{H}} \in \mathbb{C}$$

Degenerates information  
along  $t$

$x_1(t), x_2(t) \in \mathbb{C}$



$t_0 \quad t_1 \quad t_2 \quad t_3 \quad t_4 \quad t_5 \quad t_6 \quad t_7$

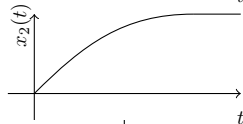
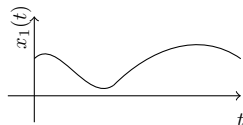
$c_0 \quad c_1 \quad c_2 \quad c_3 \quad c_4 \quad c_5 \quad c_6 \quad c_7$

$$c_i = \langle x_1(t_i), x_2(t_i) \rangle_{\mathcal{X}} \in \mathbb{C}$$

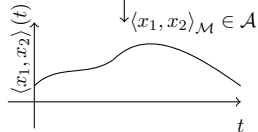
Fails to capture  
continuous behavior  
(derivatives, total variation,  
frequency components,...)

## Algorithms in RKHM

$x_1, x_2 \in \mathcal{M}$



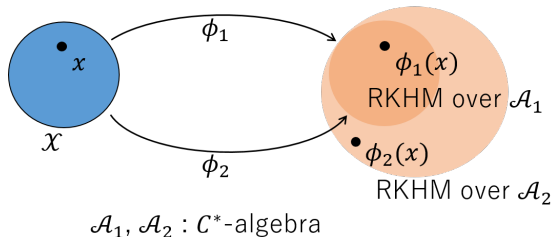
$\langle x_1, x_2 \rangle_{\mathcal{M}} \in \mathcal{A}$



Capture and control  
continuous behavior

# Advantages of RKHM

- **Enlarge representation spaces** using  $C^*$ -algebras (e.g. use the  $C^*$ -algebra of continuous functions for functional data).



- Make use of the **product structure**.  
e.g. polynomial kernel  $k(x, y) = x^*y + x^*x^*yy$  ( $x, y \in \mathcal{A}_1$  or  $\mathcal{A}_2$ )
- Use the **operator norm** to alleviate the dependency of the error on data dimension.

# Review of reproducing kernel Hilbert $C^*$ -module

$\mathcal{A}$ :  $C^*$ -algebra

RKHS ( $\mathcal{H}_k$ ):

- $\mathbb{C}$ -valued positive definite kernel  $k$
- $\mathbb{C}$ -valued functions
- $\mathbb{C}$ -valued inner product

RKHM over  $\mathcal{A}$  ( $\mathcal{M}_k$ ):

- $\mathcal{A}$ -valued positive definite kernel  $k$
- $\mathcal{A}$ -valued functions
- $\mathcal{A}$ -valued inner product

# Reproducing kernel Hilbert $C^*$ -module (RKHM)

Let  $\mathcal{X}$  be a set. A map  $k : \mathcal{X} \times \mathcal{X} \rightarrow \mathcal{A}$  is called an  **$\mathcal{A}$ -valued positive definite kernel** if it satisfies:

1.  $k(x, y) = k(y, x)^*$  for  $x, y \in \mathcal{X}$  and
2.  $\sum_{t,s=1}^n c_t^* k(x_t, x_s) c_s \geq 0$  for  $n \in \mathbb{N}$ ,  $c_1, \dots, c_n \in \mathcal{A}$ ,  $x_1, \dots, x_n \in \mathcal{X}$ .

$\phi(x) := k(\cdot, x)$  ( $\phi : \mathcal{X} \rightarrow \mathcal{A}^{\mathcal{X}}$ : feature map associated with  $k$ ),

$$\mathcal{M}_{k,0} := \left\{ \sum_{t=1}^n \phi(x_t) c_t \mid n \in \mathbb{N}, c_t \in \mathcal{A}, x_t \in \mathcal{X} \right\}. \quad (3)$$

We can define an  **$\mathcal{A}$ -valued inner product**  $\langle \cdot, \cdot \rangle_k : \mathcal{M}_{k,0} \times \mathcal{M}_{k,0} \rightarrow \mathcal{A}$  as

$$\left\langle \sum_{s=1}^n \phi(x_s) c_s, \sum_{t=1}^l \phi(y_t) d_t \right\rangle_k := \sum_{s=1}^n \sum_{t=1}^l c_s^* k(x_s, y_t) d_t. \quad (4)$$

Reproducing property:  $\langle \phi(x), v \rangle_k = v(x)$  for  $v \in \mathcal{M}_k$  and  $x \in \mathcal{X}$

**RKHM  $\mathcal{M}_k$ : completion of  $\mathcal{M}_{k,0}$**

# Representer theorem in RKHMs

To generalize complex-valued supervised problems to  $\mathcal{A}$ -valued ones, we show a representer theorem.

$\mathcal{M}_k$ : RKHM over  $\mathcal{A}$ ,  $|\cdot|_k$ : absolute value in  $\mathcal{M}_k$   
 $\mathcal{A}_+ := \{a \in \mathcal{A} \mid \exists b \in \mathcal{A} \text{ such that } a = b^*b\}$

## Theorem 2 Representer theorem in RKHMs

Let  $\mathcal{A}$  be a unital  $C^*$ -algebra,  $x_1, \dots, x_n \in \mathcal{X}$  and  $a_1, \dots, a_n \in \mathcal{A}$ . Let  $h : \mathcal{X} \times \mathcal{A}^2 \rightarrow \mathcal{A}_+$  be an error function and  $g : \mathcal{A}_+ \rightarrow \mathcal{A}_+$  satisfy  $g(c) < g(d)$  for  $c < d$ . If  $\text{Span}_{\mathcal{A}}\{\phi(x_i)\}_{i=1}^n$  is closed, any  $x \in \mathcal{M}_k$  minimizing  $f(w) := \sum_{i=1}^n h(x_i, a_i, x(x_i)) + g(|x|_k)$  admits a representation of the form  $\sum_{i=1}^n \phi(x_i)c_i$  for some  $c_1, \dots, c_n \in \mathcal{A}$ .

### Key point of the proof:

For a Hilbert  $C^*$ -module  $\mathcal{M}$  over a unital  $C^*$ -algebra  $\mathcal{A}$  and any finitely generated closed submodule  $\mathcal{V}$  of  $\mathcal{M}$ ,  $x \in \mathcal{M}$  is decomposed into  $x = x_1 + x_2$  where  $x_1 \in \mathcal{V}$  and  $x_2 \in \mathcal{V}^\perp$ .

# Approximate representer theorem in RKHMs

If  $\mathcal{A}$  is a von Neumann algebra, we can show an approximate representer theorem under mild conditions.

## Theorem 3 Approximate representer theorem in RKHMs

Let  $\mathcal{A}$  be a **von Neumann-algebra**,  $x_1, \dots, x_n \in \mathcal{X}$  and  $a_1, \dots, a_n \in \mathcal{A}$ . Let  $h : \mathcal{X} \times \mathcal{A}^2 \rightarrow \mathcal{A}_+$  be a **Lipschitz continuous** error function with Lipschitz constant  $L$  and  $g : \mathcal{A}_+ \rightarrow \mathcal{A}_+$  satisfy  $g(c) < g(d)$  for  $c < d$ . Assume  $f(x) := \sum_{i=1}^n h(x_i, a_i, x(x_i)) + g(|x|_k)$  has a minimizer  $x$ . Then, for any  $\epsilon > 0$ , there exists  $v \in \mathcal{M}_k$  of the form  $\sum_{i=1}^n \phi(x_i)c_i$  such that  $\|f(v) - f(x)\|_{\mathcal{A}} \leq Ln\epsilon\|x\|_{\mathcal{A}}$ .

### Key point of the proof:

If  $\mathcal{A}$  is a von Neumann-algebra, we can apply the Gram–Schmidt orthonormalization to construct a module approximating the module generated by  $\{\phi(x_i)\}_{i=1}^n$ .

# Conclusion

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- RKHM is a natural generalization of RKHS.
- We showed a representer theorem and an approximate representer theorem in RKHMs and in RKHMs.
- RKHMs are useful for analyzing image data and functional data.