Supplementary: Optimal Rates of Sketched-regularized Algorithms for Least-squares Regression over Hilbert Spaces

In this appendix, we first prove the lemmas stated in Section 4 and Corollary 5. We then review how the regression setting considered in this paper covers non-parametric regression with kernel methods. Finally, we construct a simple example for the non-attainable case, showing that the target function f_H is not necessarily to be in H_ρ .

A. Proofs for Lemmas in Section 4 and Corollary 5

For notational simplicity, we denote

$$\mathcal{R}_{\lambda}(u) = 1 - \mathcal{G}_{\lambda}(u)u,\tag{43}$$

and

$$\mathcal{N}(\lambda) = \operatorname{tr}(\mathcal{T}(\mathcal{T} + \lambda)^{-1}).$$

To proceed the proof, we need some basic operator inequalities.

Lemma 16. (Fujii et al., 1993) Let A and B be two positive bounded linear operators on a separable Hilbert space. Then

$$||A^s B^s|| \le ||AB||^s$$
, when $0 \le s \le 1$.

Lemma 17. Let H_1, H_2 be two separable Hilbert spaces and $S: H_1 \to H_2$ a compact operator. Then for any function $f: [0, ||S||] \to [0, \infty[$,

$$f(\mathcal{SS}^*)\mathcal{S} = \mathcal{S}f(\mathcal{S}^*\mathcal{S}).$$

Proof. The result can be proved using singular value decomposition of a compact operator.

Lemma 18. Let A and B be two non-negative bounded linear operators on a separable Hilbert space with $\max(\|A\|, \|B\|) \le \kappa^2$ for some non-negative κ^2 . Then for any $\zeta > 0$,

$$||A^{\zeta} - B^{\zeta}|| \le C_{\zeta,\kappa} ||A - B||^{\zeta \wedge 1},\tag{44}$$

where

$$C_{\zeta,\kappa} = \begin{cases} 1 & \text{when } \zeta \le 1, \\ 2\zeta\kappa^{2\zeta-2} & \text{when } \zeta > 1. \end{cases}$$
 (45)

Proof. The proof is based on the fact that u^{ζ} is operator monotone if $0 < \zeta \le 1$. While for $\zeta \ge 1$, the proof can be found in, e.g., (Dicker et al., 2016).

Lemma 19. Let X and A be bounded linear operators on a separable Hilbert space. Suppose that $X \succeq 0$ and $||A|| \leq 1$. Then for any $s \in [0, 1]$,

$$X^*A^sX \le (X^*AX)^s.$$

Proof. Following from (Hansen, 1980) and the fact that the function u^s with $s \in [0,1]$ is operator monotone.

A.1. Proof of Proposition 7

Adding and subtracting with the same term, and using the triangle inequality, we have

$$\|\mathcal{L}^{-a}(\mathcal{S}_{\varrho}\omega_{\lambda}^{\mathbf{z}} - f_{H})\|_{\varrho} \leq \|\mathcal{L}^{-a}\mathcal{S}_{\varrho}(\omega_{\lambda}^{\mathbf{z}} - \omega_{\lambda})\|_{\varrho} + \|\mathcal{L}^{-a}(\mathcal{S}_{\varrho}\omega_{\lambda} - f_{H})\|_{\varrho}.$$

Applying Part 1) of Lemma 6 to bound the last term, with $0 \le a \le \zeta$,

$$\|\mathcal{L}^{-a}(\mathcal{S}_{\rho}\omega_{\lambda}^{\mathbf{z}} - f_{H})\|_{\rho} \leq \|\mathcal{L}^{-a}\mathcal{S}_{\rho}(\omega_{\lambda}^{\mathbf{z}} - \omega_{\lambda})\|_{\rho} + R\lambda^{\zeta - a}$$
$$\leq \|\mathcal{L}^{-a}\mathcal{S}_{\rho}\mathcal{T}^{a - \frac{1}{2}}\|\|\mathcal{T}^{\frac{1}{2} - a}(\omega_{\lambda}^{\mathbf{z}} - \omega_{\lambda})\|_{H} + R\lambda^{\zeta - a}.$$

Using the spectral theorem for compact operators, $\mathcal{L} = \mathcal{S}_{\rho} \mathcal{S}_{\rho}^*$, and $\mathcal{T} = \mathcal{S}_{\rho}^* \mathcal{S}_{\rho}$, we have

$$\|\mathcal{L}^{-a}\mathcal{S}_{o}\mathcal{T}^{a-\frac{1}{2}}\| \leq 1,$$

and thus

$$\|\mathcal{L}^{-a}(\mathcal{S}_{\varrho}\omega_{\lambda}^{\mathbf{z}} - f_{H})\|_{\varrho} \leq \|\mathcal{T}^{\frac{1}{2}-a}(\omega_{\lambda}^{\mathbf{z}} - \omega_{\lambda})\|_{H} + R\lambda^{\zeta-a}.$$

Adding and subtracting with the same term, and using the triangle inequality,

$$\|\mathcal{L}^{-a}(\mathcal{S}_{\rho}\omega_{\lambda}^{\mathbf{z}} - f_{H})\|_{\rho} \leq \|\mathcal{T}^{\frac{1}{2}-a}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} + \|\mathcal{T}^{\frac{1}{2}-a}(I - P)\omega_{\lambda}\|_{H} + R\lambda^{\zeta - a}.$$

Since P is an orthogonal projected operator and $a \in [0, \frac{1}{2}]$, we have

$$\begin{split} & \|\mathcal{T}^{\frac{1}{2}-a}(I-P)\omega_{\lambda}\|_{H} \\ = & \|\mathcal{T}^{\frac{1}{2}(1-2a)}(I-P)^{1-2a}(I-P)\omega_{\lambda}\|_{H} \\ \leq & \|\mathcal{T}^{\frac{1}{2}(1-2a)}(I-P)^{1-2a}\|\|(I-P)\mathcal{T}^{\frac{1}{2}}\|\|\mathcal{T}^{-\frac{1}{2}}\omega_{\lambda}\|_{H} \\ \leq & \|\mathcal{T}^{\frac{1}{2}}(I-P)\|^{1-2a}\|(I-P)\mathcal{T}^{\frac{1}{2}}\|\tau R\kappa^{2(\zeta-1)+}\lambda^{(\zeta-1)-2} \\ = & \Delta_{5}^{1-a}\tau R\kappa^{2(\zeta-1)+}\lambda^{(\zeta-1)-2}, \end{split}$$

(where for the last second inequality, we used Lemma 16 and Part 2) of Lemma 6), and we subsequently get that

$$\|\mathcal{L}^{-a}(\mathcal{S}_{\rho}\omega_{\lambda}^{\mathbf{z}} - f_{H})\|_{\rho} \leq \|\mathcal{T}^{\frac{1}{2}-a}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} + \tau R\kappa^{2(\zeta-1)} + \lambda^{(\zeta-1)} - \Delta_{5}^{1-a} + R\lambda^{\zeta-a}.$$

Since for all $\omega \in H$, and $a \in [0, \frac{1}{2}]$,

$$\begin{split} \|\mathcal{T}^{\frac{1}{2}-a}\omega\|_{H} &\leq & \|\mathcal{T}_{\lambda}^{\frac{1}{2}-a}\mathcal{T}_{\mathbf{x}\lambda}^{a-\frac{1}{2}}\|\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}-a}\omega\|_{H} \\ &\leq & \lambda^{-a}\|\mathcal{T}_{\lambda}^{\frac{1}{2}-a}\mathcal{T}_{\mathbf{x}\lambda}^{a-\frac{1}{2}}\|\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\omega\|_{H} \\ &\leq & \lambda^{-a}\|\mathcal{T}_{\lambda}^{\frac{1}{2}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\|^{1-2a}\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\omega\|_{H} \\ &\leq & \lambda^{-a}\Delta_{1}^{\frac{1}{2}-a}\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\omega\|_{H} \end{split}$$

(where we used Lemma 16 for the last second inequality), we get

$$\|\mathcal{L}^{-a}(\mathcal{S}_{\rho}\omega_{\lambda}^{\mathbf{z}} - f_{H})\|_{\rho} \leq \lambda^{-a}\Delta_{1}^{\frac{1}{2}-a}\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} + \tau R\kappa^{2(\zeta-1)} + \lambda^{(\zeta-1)}\Delta_{5}^{1-a} + R\lambda^{\zeta-a}. \tag{46}$$

In what follows, we estimate $\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H}$.

Introducing with (11), with $P^2 = P$,

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} = \|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}P(\mathcal{G}_{\lambda}(P\mathcal{T}_{\mathbf{x}}P)P\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - P\omega_{\lambda})\|_{H}.$$

Since for any $\omega \in H$,

$$\|\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}} P \omega\|_{H}^{2} = \langle P \mathcal{T}_{\mathbf{x}} \lambda P \omega, \omega \rangle_{H} \leq \langle (P \mathcal{T}_{\mathbf{x}} P + \lambda) \omega, \omega \rangle_{H} = \|(P \mathcal{T}_{\mathbf{x}} P + \lambda)^{\frac{1}{2}} \omega\|_{H}^{2}.$$

and we thus get

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} \leq \|\mathcal{U}_{\lambda}^{\frac{1}{2}}(\mathcal{G}_{\lambda}(\mathcal{U})P\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - P\omega_{\lambda})\|_{H},$$

where we denote

$$\mathcal{U} = P\mathcal{T}_{\mathbf{x}}P, \quad \mathcal{U}_{\lambda} = \mathcal{U} + \lambda.$$
 (47)

Subtracting and adding with the same term, and applying the triangle inequality, with the notation \mathcal{R}_{λ} given by (43) and $P^2 = P$, we have

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} \leq \|\underbrace{\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{G}_{\lambda}(\mathcal{U})P(\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - \mathcal{T}_{\mathbf{x}}P\omega_{\lambda})}_{\mathbf{Term},\mathbf{A}}\|_{H} + \|\underbrace{\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\omega_{\lambda}}_{\mathbf{Term},\mathbf{B}}\|_{H}. \tag{48}$$

We will estimate the above two terms of the right-hand side.

Estimating $\|\mathbf{Term.A}\|_H$:

Note that

$$\begin{split} &(\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{G}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}})(\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{G}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}})^{*}\\ &=\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{G}_{\lambda}(\mathcal{U})(\mathcal{U}+\lambda P^{2})\mathcal{G}_{\lambda}(\mathcal{U})\mathcal{U}_{\lambda}^{\frac{1}{2}}\\ &\preceq\left[\mathcal{U}_{\lambda}\mathcal{G}_{\lambda}(\mathcal{U})\right]^{2}, \end{split}$$

where we used $P^2 = P \leq I$ for the last inequality. Thus, combing with $||A|| = ||A^*A||^{\frac{1}{2}}$.

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{G}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\| \leq \|\mathcal{U}_{\lambda}\mathcal{G}_{\lambda}(\mathcal{U})\|.$$

Using the spectral theorem, with $\|\mathcal{U}\| \leq \|\mathcal{T}_{\mathbf{x}}\| \leq \kappa^2$ (implied by (6)), and then applying (12),

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{G}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\| \leq \sup_{u \in [0,\kappa^{2}]} |(u+\lambda)\mathcal{G}_{\lambda}(u)| \leq \tau.$$

Using the above inequality, and by a simple calculation,

$$\|\mathbf{Term}.\mathbf{A}\|_{H} \leq \|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{G}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\|\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}(\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - \mathcal{T}_{\mathbf{x}}P\omega_{\lambda})\| \leq \tau\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}(\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - \mathcal{T}_{\mathbf{x}}P\omega_{\lambda})\|.$$

Adding and subtracting with the same terms, and using the triangle inequality,

$$\begin{split} \|\mathbf{Term.A}\|_{H} &\leq \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}(\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - \mathcal{T}_{\mathbf{x}}\omega_{\lambda})\|_{H} + \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I - P)\omega_{\lambda})\|_{H} \\ &\leq \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\lambda}^{\frac{1}{2}}\|\|\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - \mathcal{T}_{\mathbf{x}}\omega_{\lambda})\|_{H} + \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I - P)\omega_{\lambda})\|_{H} \\ &\leq \tau \Delta_{1}^{\frac{1}{2}}\|\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{S}_{\mathbf{x}}^{*}\mathbf{y} - \mathcal{T}_{\mathbf{x}}\omega_{\lambda})\|_{H} + \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I - P)\omega_{\lambda})\|_{H} \\ &\leq \tau \Delta_{1}^{\frac{1}{2}}(\Delta_{2} + \|\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{T}\omega_{\lambda} - \mathcal{S}_{\rho}^{*}f_{H})\|_{H}) + \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I - P)\omega_{\lambda})\|_{H} \\ &\leq \tau \Delta_{1}^{\frac{1}{2}}(\Delta_{2} + \|\mathcal{T}_{\lambda}^{-\frac{1}{2}}\mathcal{S}_{\rho}^{*}\|\|\mathcal{S}_{\rho}\omega_{\lambda} - f_{H}\|_{\rho}) + \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I - P)\omega_{\lambda})\|_{H}, \end{split}$$

where we used $\mathcal{T} = \mathcal{S}_{\rho}^* \mathcal{S}_{\rho}$ for the last inequality. Applying Part 1) of Lemma 6 and $\|\mathcal{T}_{\lambda}^{-\frac{1}{2}} \mathcal{S}_{\rho}^*\| \leq 1$,

$$\|\mathbf{Term.A}\|_{H} \le \tau \Delta_{1}^{\frac{1}{2}} (\Delta_{2} + R\lambda^{\zeta}) + \tau \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}} \mathcal{T}_{\mathbf{x}} (I - P)\omega_{\lambda})\|_{H}. \tag{49}$$

In what follows, we estimate $\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda}\|_{H}$, considering two different cases. Case $\zeta \leq 1$.

We have

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda}\|_{H} \leq \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\|\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\mathcal{T}_{\lambda}\|\|\mathcal{T}_{\lambda}^{\frac{1}{2}}(I-P)\omega_{\lambda}\|_{H} \leq \Delta_{1}^{\frac{1}{2}}\|\mathcal{T}_{\lambda}^{\frac{1}{2}}(I-P)\omega_{\lambda}\|_{H}.$$

Since P is a projection operator, $(I - P)^2 = I - P$, and we thus have

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda}\|_{H} \leq \Delta_{1}^{\frac{1}{2}}\|\mathcal{T}_{\lambda}^{\frac{1}{2}}(I-P)\|\|(I-P)\mathcal{T}^{\frac{1}{2}}\|\|\mathcal{T}^{-\frac{1}{2}}\omega_{\lambda}\|_{H} \leq \tau\Delta_{1}^{\frac{1}{2}}\|\mathcal{T}_{\lambda}^{\frac{1}{2}}(I-P)\|\Delta_{5}^{\frac{1}{2}}R\lambda^{\zeta-1},$$

where for the last inequality, we used Part 2) of Lemma 6. Note that for any $\omega \in H$ with $\|\omega\|_H = 1$,

$$\|\mathcal{T}_{\lambda}^{\frac{1}{2}}(I-P)\omega\|_{H}^{2} = \langle \mathcal{T}_{\lambda}(I-P)\omega, (I-P)\omega \rangle_{H} = \|\mathcal{T}^{\frac{1}{2}}(I-P)\omega\|_{H}^{2} + \lambda \|(I-P)\omega\|_{H}^{2} \leq \|\mathcal{T}^{\frac{1}{2}}(I-P)\|^{2} + \lambda \leq \Delta_{5} + \lambda.$$

It thus follows that

$$\|\mathcal{T}_{\lambda}^{\frac{1}{2}}(I-P)\|_{H} \le (\Delta_{5}+\lambda)^{\frac{1}{2}},$$
 (50)

and thus

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda}\|_{H} \leq \Delta_{1}^{\frac{1}{2}}(\Delta_{5}+\lambda)\tau R\lambda^{\zeta-1}.$$

Introducing the above into (49), we know that **Term.A** can be estimated as ($\zeta \leq 1$)

$$\|\mathbf{Term.A}\|_{H} \le \tau \Delta_{1}^{\frac{1}{2}} \left(\Delta_{2} + (\tau + 1)R\lambda^{\zeta} + \tau R\lambda^{\zeta - 1}\Delta_{5} \right). \tag{51}$$

Case $\zeta \geq 1$.

We first have

$$\begin{split} \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda}\|_{H} &\leq \Delta_{1}^{\frac{1}{2}}\|\mathcal{T}_{\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda})\|_{H} \\ &\leq \Delta_{1}^{\frac{1}{2}}\left(\|\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{T}_{\mathbf{x}}-\mathcal{T})(I-P)\omega_{\lambda}\|_{H} + \|\mathcal{T}_{\lambda}^{-\frac{1}{2}}\mathcal{T}(I-P)\omega_{\lambda}\|_{H}\right) \\ &\leq \Delta_{1}^{\frac{1}{2}}\left(\Delta_{4}\|(I-P)\omega_{\lambda}\|_{H} + \|\mathcal{T}^{\frac{1}{2}}(I-P)\omega_{\lambda}\|_{H}\right). \end{split}$$

Since P is a projection operator, $(I - P)^2 = I - P$, we thus have

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda}\|_{H} \leq \Delta_{1}^{\frac{1}{2}} \left(\Delta_{4}\|I-P\|\|\mathcal{T}^{\frac{1}{2}}\|\|\mathcal{T}^{-\frac{1}{2}}\omega_{\lambda}\|_{H} + \|\mathcal{T}^{\frac{1}{2}}(I-P)\|\|(I-P)\mathcal{T}^{\frac{1}{2}}\|\|\mathcal{T}^{-\frac{1}{2}}\omega_{\lambda}\|_{H}\right)$$

$$\leq \Delta_{1}^{\frac{1}{2}} \left(\kappa\Delta_{4} + \Delta_{5}\right)\|\mathcal{T}^{-\frac{1}{2}}\omega_{\lambda}\|_{H},$$

where we used (3) for the last inequality. Applying Part 2) of Lemma 6, we get

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}(I-P)\omega_{\lambda}\|_{H} \leq \Delta_{1}^{\frac{1}{2}}\left(\kappa\Delta_{4}+\Delta_{5}\right)\tau\kappa^{2(\zeta-1)}R.$$

Introducing the above into (49), we get for $\zeta \geq 1$,

$$\|\mathbf{Term.A}\|_{H} \le \tau \Delta_{1}^{\frac{1}{2}} \left(\Delta_{2} + R\lambda^{\zeta} + (\kappa \Delta_{4} + \Delta_{5}) \tau \kappa^{2(\zeta - 1)} R \right). \tag{52}$$

Estimating $\|\mathbf{Term.B}\|_{H}$:

We estimate $\|\mathbf{Term.B}\|_{H}$, considering two different cases.

Case I: $\zeta \leq 1$.

We first have

$$\begin{aligned} \mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}})^{*} &= & \mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})(\mathcal{U} + \lambda P^{2})\mathcal{R}_{\lambda}(\mathcal{U})\mathcal{U}_{\lambda}^{\frac{1}{2}} \\ &\leq & \left(\mathcal{R}_{\lambda}(\mathcal{U})\mathcal{U}_{\lambda}\right)^{2}, \end{aligned}$$

where we used $P^2=P\preceq I$ for the last inequality. Thus, according to $\|A\|=\|AA^*\|^{\frac{1}{2}},$

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\| \leq \|\mathcal{R}_{\lambda}(\mathcal{U})\mathcal{U}_{\lambda}\|.$$

Using the spectral theorem and (13), and noting that $\|\mathcal{U}\| \leq \|P\|^2 \|\mathcal{T}_{\mathbf{x}}\| \leq \kappa^2$ by (6), we get

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\| \leq \sup_{u \in [0,\kappa^{2}]} |\mathcal{R}_{\lambda}(u)(u+\lambda)| \leq \lambda.$$

Using the above inequality and by a direct calculation,

$$\|\mathbf{Term.B}\|_{H} \leq \|\mathcal{U}_{\lambda}^{\frac{1}{2}} \mathcal{R}_{\lambda}(\mathcal{U}) P \mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}} \|\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}} \mathcal{T}_{\lambda}^{\frac{1}{2}} \|\|\mathcal{T}^{-\frac{1}{2}} \omega_{\lambda}\|_{H} \leq \lambda \Delta_{1}^{\frac{1}{2}} \|\mathcal{T}^{-\frac{1}{2}} \omega_{\lambda}\|_{H}$$

Applying Part 2) of Lemma 6, we get

$$\|\mathbf{Term.B}\|_{H} \le \tau R \lambda^{\zeta} \Delta_{1}^{\frac{1}{2}}.$$
 (53)

Applying the above and (51) into (48), we know that for any $\zeta \in [0, 1]$,

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} \leq \tau \Delta_{1}^{\frac{1}{2}} \left(\Delta_{2} + (2\tau + 1)R\lambda^{\zeta} + \tau R\Delta_{5}\lambda^{\zeta - 1}\right).$$

Using the above into (46), we can prove the first desired result.

Case II: $\zeta \geq 1$

We denote

$$\mathcal{V} = \mathcal{T}_{\mathbf{x}}^{\frac{1}{2}} P \mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}, \quad \mathcal{V}_{\lambda} = \mathcal{V} + \lambda. \tag{54}$$

Noting that $\mathcal{U}=P\mathcal{T}_{\mathbf{x}}P=P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}(P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}})^*$, thus following from Lemma 17 (with $f(u)=(u+\lambda)^{\frac{1}{2}}\mathcal{R}_{\lambda}(u)$) and $P^2=P$,

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}}^{\zeta-\frac{1}{2}}\| = \|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})(P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}})\mathcal{T}_{\mathbf{x}}^{\zeta-1}\| = \|(P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}})\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})\mathcal{T}_{\mathbf{x}}^{\zeta-1}\|.$$

Adding and subtracting with the same term, using the triangle inequality

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}}^{\zeta-\frac{1}{2}}\| \leq \|P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})\mathcal{V}^{\zeta-1}\| + \|P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})(\mathcal{T}_{\mathbf{x}}^{\zeta-1} - \mathcal{V}^{\zeta-1})\|$$

$$\leq \|P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})\mathcal{V}^{\zeta-1}\| + \|P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})\|\|\mathcal{T}_{\mathbf{x}}^{\zeta-1} - \mathcal{V}^{\zeta-1}\|.$$

Using Lemma 18, with (6) and $\|\mathcal{V}\| \leq \|\mathcal{T}_{\mathbf{x}}\| \leq \kappa^2$,

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}}^{\zeta-\frac{1}{2}}\| \leq \|P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})\mathcal{V}^{\zeta-1}\| + \|P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})\|\kappa^{2(\zeta-2)_{+}}\|\mathcal{T}_{\mathbf{x}} - \mathcal{V}\|^{(\zeta-1)\wedge 1}.$$

Using $||A|| = ||A^*A||^{\frac{1}{2}}$, $P^2 = P$, the spectral theorem, and (13), for any $s \in [1, \tau]$,

$$\begin{split} \|P\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\mathcal{V}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{V})\mathcal{V}^{s-1}\| = &\|\mathcal{V}^{s-1}\mathcal{R}_{\lambda}(\mathcal{V})\mathcal{V}_{\lambda}\mathcal{V}\mathcal{R}_{\lambda}(\mathcal{V})\mathcal{V}^{s-1}\|^{\frac{1}{2}} \\ \leq \sup_{u \in [0,\kappa^{2}]} |\mathcal{R}_{\lambda}(u)u^{s-\frac{1}{2}}(u+\lambda)^{\frac{1}{2}}| \leq \lambda^{s}, \end{split}$$

and thus we get

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}-a}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}}^{\zeta-\frac{1}{2}}\| \leq \lambda^{\zeta} + \lambda \kappa^{2(\zeta-2)_{+}}\|\mathcal{T}_{\mathbf{x}} - \mathcal{V}\|^{(\zeta-1)\wedge 1}.$$

Using Lemma 14, $(I - P)^2 = I - P$ and $||A^*A|| = ||A||^2$, we have

$$\|\mathcal{T}_{\mathbf{x}} - \mathcal{V}\| = \|\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}(I - P)\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\| \le \|\mathcal{T}_{\mathbf{x}} - \mathcal{T}\| + \|\mathcal{T}^{\frac{1}{2}}(I - P)\mathcal{T}^{\frac{1}{2}}\| \le \Delta_3 + \Delta_5,$$

and we thus get

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}_{\mathbf{x}}^{\zeta-\frac{1}{2}}\| \leq \lambda^{\zeta} + \lambda \kappa^{2(\zeta-2)_{+}} (\Delta_{3} + \Delta_{5})^{(\zeta-1)\wedge 1}. \tag{55}$$

Now we are ready to estimate $\|\mathbf{Term.B}\|_{H}$. By some direct calculations and Part 2) of Lemma 6,

$$\|\mathbf{Term.B}\|_{H} \leq \|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}^{\zeta-\frac{1}{2}}\|\|\mathcal{T}^{\frac{1}{2}-\zeta}\omega_{\lambda}\|_{H} \leq \|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})P\mathcal{T}^{\zeta-\frac{1}{2}}\|\tau R.$$

Adding and subtracting with the same term, and using the triangle inequality,

$$\|\mathbf{Term.B}\|_{H} \leq \tau R \left(\|\mathcal{U}_{\lambda}^{\frac{1}{2}} \mathcal{R}_{\lambda}(\mathcal{U}) P \mathcal{T}_{\mathbf{x}}^{\zeta - \frac{1}{2}} \| + \|\mathcal{U}_{\lambda}^{\frac{1}{2}} \mathcal{R}_{\lambda}(\mathcal{U}) \| \|\mathcal{T}^{\zeta - \frac{1}{2}} - \mathcal{T}_{\mathbf{x}}^{\zeta - \frac{1}{2}} \| \right).$$

Using the spectral theorem, with $\|\mathcal{U}\| \leq \|\mathcal{T}_{\mathbf{x}}\| \leq \kappa^2$ by (6) and (13),

$$\|\mathcal{U}_{\lambda}^{\frac{1}{2}}\mathcal{R}_{\lambda}(\mathcal{U})\| = \sup_{u \in [0,\kappa^2]} |\mathcal{R}_{\lambda}(u)(u+\lambda)^{\frac{1}{2}}| \leq \lambda^{\frac{1}{2}},$$

and we thus get

$$\|\mathbf{Term.B}\|_{H} \leq \tau R \left(\|\mathcal{U}_{\lambda}^{\frac{1}{2}} \mathcal{R}_{\lambda}(\mathcal{U}) P \mathcal{T}_{\mathbf{x}}^{\zeta - \frac{1}{2}} \| + \lambda^{\frac{1}{2}} \|\mathcal{T}^{\zeta - \frac{1}{2}} - \mathcal{T}_{\mathbf{x}}^{\zeta - \frac{1}{2}} \| \right).$$

Applying Lemma 18, with (3) and (6),

$$\|\mathbf{Term.B}\|_{H} \leq \tau R \left(\|\mathcal{U}_{\lambda}^{\frac{1}{2}} \mathcal{R}_{\lambda}(\mathcal{U}) P \mathcal{T}_{\mathbf{x}}^{\zeta - \frac{1}{2}} \| + \lambda^{\frac{1}{2}} \kappa^{(2\zeta - 3)} \Delta_{3}^{(\zeta - \frac{1}{2}) \wedge 1} \right).$$

Introducing with (55),

$$\|\mathbf{Term.B}\|_{H} \leq \tau R \left(\lambda^{\zeta} + \kappa^{2(\zeta-2)} + \lambda(\Delta_{3} + \Delta_{5})^{(\zeta-1)\wedge 1} + \kappa^{(2\zeta-3)} + \lambda^{\frac{1}{2}} \Delta_{3}^{(\zeta-\frac{1}{2})\wedge 1}\right).$$

Introducing the above inequality and (52) into (48), noting that $\Delta_1 \geq 1$ and $\kappa^2 \geq 1$, we know that for any $\zeta \geq 1$,

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(\omega_{\lambda}^{\mathbf{z}} - P\omega_{\lambda})\|_{H} \leq \tau \Delta_{1}^{\frac{1}{2}} \left(\Delta_{2} + 2R\lambda^{\zeta} + \kappa^{2(\zeta-1)}R(\kappa\tau\Delta_{4} + \tau\Delta_{5} + \lambda(\Delta_{3} + \Delta_{5})^{(\zeta-1)\wedge 1} + \lambda^{\frac{1}{2}}\Delta_{3}^{(\zeta-\frac{1}{2})\wedge 1})\right).$$

Using the above into (46), and by a simple calculation, we can prove the second desired result.

A.2. Proofs of Lemma 12

We first introduce the following basic probabilistic estimate.

Lemma 20. Let $\mathcal{X}_1, \dots, \mathcal{X}_m$ be a sequence of independently and identically distributed self-adjoint Hilbert-Schmidt operators on a separable Hilbert space. Assume that $\mathbb{E}[\mathcal{X}_1] = 0$, and $\|\mathcal{X}_1\| \leq B$ almost surely for some B > 0. Let \mathcal{V} be a positive trace-class operator such that $\mathbb{E}[\mathcal{X}_1^2] \leq \mathcal{V}$. Then with probability at least $1 - \delta$, $(\delta \in]0, 1[$), there holds

$$\left\| \frac{1}{m} \sum_{i=1}^{m} \mathcal{X}_i \right\| \le \frac{2B\beta}{3m} + \sqrt{\frac{2\|\mathcal{V}\|\beta}{m}}, \qquad \beta = \log \frac{4 \operatorname{tr} \mathcal{V}}{\|\mathcal{V}\|\delta}.$$

The above lemma was first proved in (Hsu et al., 2014; Tropp, 2012) for the matrix case, and it was later extended to the general operator case in (Minsker, 2011), see also (Rudi et al., 2015; Bach, 2015; Dicker et al., 2017). We refer to (Rudi et al., 2015; Dicker et al., 2017) for the proof.

Using the above lemma, we can prove Lemma 12.

Proof of Lemma 12. We use Lemma 20 to prove the result. Let $W = m^{-\frac{1}{2}} \mathbf{G} \mathcal{S}_{\mathbf{x}}$. Denote the *i*-th row of \mathbf{G} by \mathbf{a}_i^* for all $i \in [m]$. Using $\mathcal{T}_{\mathbf{x}} = \mathcal{S}_{\mathbf{x}}^* \mathcal{S}_{\mathbf{x}}$, we have

$$\mathcal{T}_{\mathbf{x}\boldsymbol{\lambda}}^{-\frac{1}{2}}(\mathcal{T}_{\mathbf{x}} - W^*W)\mathcal{T}_{\mathbf{x}\boldsymbol{\lambda}}^{-\frac{1}{2}} = \mathcal{T}_{\mathbf{x}\boldsymbol{\lambda}}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^*(I - m^{-1}\mathbf{G}^*\mathbf{G})\mathcal{S}_{\mathbf{x}}\mathcal{T}_{\mathbf{x}\boldsymbol{\lambda}}^{-\frac{1}{2}} = \frac{1}{m}\sum_{i=1}^m \mathcal{X}_i,$$

where we let

$$\mathcal{X}_i = \mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}} \mathcal{S}_{\mathbf{x}}^* (I - \mathbf{a}_i \mathbf{a}_i^*) \mathcal{S}_{\mathbf{x}} \mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}.$$

Since $\mathbf{a}_1 \sim F$, according to the isotropy property (26) of F,

$$\mathbb{E}[\mathcal{X}_1] = \mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}} \mathcal{S}_{\mathbf{x}}^* (I - \mathbb{E}[\mathbf{a}_i \mathbf{a}_i^*]) \mathcal{S}_{\mathbf{x}} \mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}} = 0.$$

Note that

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^*\mathbf{a}_1\|_{H} = \frac{1}{n} \left\| \sum_{j=1}^{n} \mathbf{a}_1(j) \mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}} x_j \right\|_{H} \le \frac{1}{n} \sum_{j=1}^{n} |\mathbf{a}_1(j)| \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}} x_j\|_{H}.$$

Using Cauchy-Schwarz inequality and the bounded assumption (27),

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^*\mathbf{a}_1\|_H \leq \frac{1}{n}\|\mathbf{a}_1\|_2 \left(\sum_{j=1}^n \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}x_j\|_H^2\right)^{\frac{1}{2}} \leq \left(\frac{1}{n}\sum_{j=1}^n \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}x_j\|_H^2\right)^{\frac{1}{2}}.$$

According to $\operatorname{tr}(x \otimes x) = \|x\|_H^2$ and the definition of $\mathcal{T}_{\mathbf{x}}$, we know that the left-hand side is $\sqrt{\operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}})}$, and thus

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^*\mathbf{a}_1\|_H \leq \sqrt{\operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}})}.$$

Therefore.

$$\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^{*}\mathbf{a}_{1}\mathbf{a}_{1}^{*}\mathcal{S}_{\mathbf{x}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\| \leq \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^{*}\mathbf{a}_{1}\mathbf{a}_{1}^{*}\mathcal{S}_{\mathbf{x}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}) \leq \|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^{*}\mathbf{a}_{1}\|_{H}^{2} \leq \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}),$$

and by $||a - \mathbb{E}[a]|| \le ||a|| + \mathbb{E}||a||$,

$$\|\mathcal{X}_1\| \leq 2\operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}).$$

Moreover, using $\mathbb{E}[a - \mathbb{E}[a]]^2 \leq \mathbb{E}a^2$,

$$\begin{split} \mathbb{E}[\mathcal{X}_{1}^{2}] & \preceq \mathbb{E}[\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^{*}\mathbf{a}_{1}\mathbf{a}_{1}^{*}\mathcal{S}_{\mathbf{x}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}]^{2} = & \mathbb{E}[\|\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^{*}\mathbf{a}_{1}\|_{H}^{2}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^{*}\mathbf{a}_{1}\mathbf{a}_{1}^{*}\mathcal{S}_{\mathbf{x}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}] \\ & \preceq \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}})\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{S}_{\mathbf{x}}^{*}\mathbb{E}[\mathbf{a}_{1}\mathbf{a}_{1}^{*}]\mathcal{S}_{\mathbf{x}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}] \\ & = \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}})\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}. \end{split}$$

Letting $\mathcal{V} = \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}})\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}$, a simple calculation shows that

$$\|\mathcal{V}\| = \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}})\|\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}\| \leq \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}).$$

Also, $\|\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}\| = \frac{\|\mathcal{T}_{\mathbf{x}}\|}{\|\mathcal{T}_{\mathbf{x}}\| + \lambda}$

$$\frac{\operatorname{tr}(\mathcal{V})}{\|\mathcal{V}\|} = \frac{\operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}})}{\|\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}\|} = \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}) \left(1 + \frac{\lambda}{\|\mathcal{T}_{\mathbf{x}}\|}\right).$$

Applying Lemma 20, one can prove the desired result.

A.3. Proof of Lemma 13

If $\lambda \geq ||\mathcal{T}_{\mathbf{x}}||$, then the result follows trivially,

$$\|(I-P)\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\|^{2} \le \|(I-P)\|^{2}\|\mathcal{T}_{\mathbf{x}}\| \le \frac{1}{n^{\theta}}.$$

We thus only need to consider the case $\lambda \leq \|\mathcal{T}_{\mathbf{x}}\|$. Let $M = m^{-1}\mathcal{S}_{\mathbf{x}}^*\mathbf{G}^*\mathbf{G}\mathcal{S}_{\mathbf{x}}$ and $M_{\lambda} = M + \lambda I$. Applying Lemma 12, we know that there exists a subset $U_{\mathbf{x}}$ of $\mathbb{R}^{m \times n}$ with measure at least $1 - \delta$, such that

$$\left\| \mathcal{T}_{\mathbf{x}\lambda}^{-1/2} (\mathcal{T}_{\mathbf{x}} - M) \mathcal{T}_{\mathbf{x}\lambda}^{-1/2} \right\| \le \frac{4\mathcal{N}_{\mathbf{x}}(\lambda)\beta}{3m} + \sqrt{\frac{2\mathcal{N}_{\mathbf{x}}(\lambda)\beta}{m}}, \quad \forall \mathbf{G} \in U_{\mathbf{x}}.$$
 (56)

Using Condition (39),

$$\mathcal{N}_{\mathbf{x}}(\lambda) \leq b_{\gamma} n^{\theta \gamma}.$$

With $\lambda \leq \|\mathcal{T}_{\mathbf{x}}\|$, we have

$$\beta \le \log \frac{4b_{\gamma}n^{\theta\gamma}(1+\lambda/\|\mathcal{T}_{\mathbf{x}}\|)}{\delta} \le \log \frac{8b_{\gamma}n^{\theta\gamma}}{\delta},$$

and, combining with (40),

$$\frac{4\mathcal{N}_{\mathbf{x}}(\lambda)\beta}{3m} + \sqrt{\frac{2\mathcal{N}_{\mathbf{x}}(\lambda)\beta}{m}} \le \frac{2}{3}.$$

Thus,

$$\left\| \mathcal{T}_{\mathbf{x}\lambda}^{-1/2} (\mathcal{T} - M) \mathcal{T}_{\mathbf{x}\lambda}^{-1/2} \right\| \le \frac{2}{3}, \quad \forall \mathbf{G} \in U_{\mathbf{x}}.$$

Following from (Caponnetto & De Vito, 2007),

$$\|M_{\lambda}^{-1/2}\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\|^{2} = \|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}M_{\lambda}^{-1}\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\|^{2} = \|(I - \mathcal{T}_{\mathbf{x}\lambda}^{-1/2}(\mathcal{T}_{\mathbf{x}} - M)\mathcal{T}_{\mathbf{x}\lambda}^{-1/2})^{-1/2}\|,$$

we get

$$\|M_{\lambda'}^{-1/2}\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\|^2 \le 3, \quad \forall \mathbf{G} \in U_{\mathbf{x}}. \tag{57}$$

Let $W = m^{-1/2} \mathbf{G} \mathcal{S}_{\mathbf{x}}$. As P is the projection operator onto $\overline{range\{W^*\}}$,

$$P = W^*(WW^*)^{\dagger}W \succeq W^*(WW^* + \lambda)^{-1}W = W^*W(W^*W + \lambda)^{-1} = M(M + \lambda)^{-1},$$

where for the last second equality, we used Lemma 17. Thus (Rudi et al., 2015),

$$I - P \prec I - M(M + \lambda)^{-1} = \lambda (M + \lambda)^{-1}.$$

It thus follows that

$$\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}(I-P)\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}} \leq \lambda \mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}(M+\lambda)^{-1}\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}} \leq \lambda \mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(M+\lambda)^{-1}\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}.$$

Using $||A^*A||^2 = ||A||^2$ and the above,

$$\|(I-P)\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\|^{2} = \|\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}(I-P)\mathcal{T}_{\mathbf{x}}^{\frac{1}{2}}\| \le \lambda \|\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}(M+\lambda)^{-1}\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\| = \lambda \|(M+\lambda)^{-1/2}\mathcal{T}_{\mathbf{x}\lambda}^{\frac{1}{2}}\|^{2}.$$
 (58)

Applying (57), one can prove the desired result.

A.4. Proof of Lemma 14

Since P is a projection operator, $(I - P)^2 = I - P$. Then

$$||A^{s}(I-P)A^{t}|| = ||A^{s}(I-P)(I-P)A^{t}|| \le ||A^{s}(I-P)||||(I-P)A^{t}||.$$

Moreover, by Lemma 16,

$$||A^{s}(I-P)|| = ||A^{\frac{1}{2}2s}(I-P)^{2s}|| \le ||A^{\frac{1}{2}}(I-P)||^{2s}.$$

Similarly, $||(I-P)A^t|| \le ||(I-P)A^{\frac{1}{2}}||^{2t}$. Thus,

$$||A^{s}(I-P)A^{t}|| \le ||A^{\frac{1}{2}}(I-P)||^{2s}||(I-P)A^{\frac{1}{2}}||^{2t} = ||(I-P)A^{\frac{1}{2}}||^{2(t+s)}.$$

Using $||D||^2 = ||D^*D||$,

$$||A^{s}(I-P)A^{t}|| \le ||(I-P)A(I-P)||^{t+s}.$$

Adding and subtracting with the same term, using the triangle inequality, and noting that $||I - P|| \le 1$ and $s + t \le 1$,

$$||A^{s}(I-P)A^{t}|| \le ||(I-P)A(I-P)||^{t+s}$$

$$\le (||(I-P)(A-B)(I-P)|| + ||(I-P)B(I-P)||)^{t+s}$$

$$\le ||A-B||^{s+t} + ||(I-P)B(I-P)||^{s+t},$$

which leads to the desired result using $||D^*D|| = ||DD^*||$.

A.5. Proof of Lemma 15

To prove the result, we need the following concentration inequality.

Lemma 21. Let w_1, \dots, w_m be i.i.d random variables in a separable Hilbert space with norm $\|\cdot\|$. Suppose that there are two positive constants B and σ^2 such that

$$\mathbb{E}[\|w_1 - \mathbb{E}[w_1]\|^l] \le \frac{1}{2} l! B^{l-2} \sigma^2, \quad \forall l \ge 2.$$
 (59)

Then for any $0 < \delta < 1/2$, the following holds with probability at least $1 - \delta$,

$$\left\| \frac{1}{m} \sum_{k=1}^{m} w_m - \mathbb{E}[w_1] \right\| \le 2 \left(\frac{B}{m} + \frac{\sigma}{\sqrt{m}} \right) \log \frac{2}{\delta}.$$

In particular, (59) holds if

$$||w_1|| \le B/2 \text{ a.s.}, \quad and \quad \mathbb{E}[||w_1||^2] \le \sigma^2.$$
 (60)

The above lemma is a reformulation of the concentration inequality for sums of Hilbert-space-valued random variables from (Pinelis & Sakhanenko, 1986). We refer to (Smale & Zhou, 2007; Caponnetto & De Vito, 2007) for the detailed proof.

Proof of Lemma 15. We first use Lemma 21 to estimate $\operatorname{tr}(\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{T}_{\mathbf{x}}-\mathcal{T})\mathcal{T}_{\lambda}^{-\frac{1}{2}})$. Note that

$$\operatorname{tr}(\mathcal{T}_{\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}\mathcal{T}_{\lambda}^{-\frac{1}{2}}) = \frac{1}{n}\sum_{j=1}^{n} \|\mathcal{T}_{\lambda}^{-\frac{1}{2}}x_{j}\|_{H}^{2} = \frac{1}{n}\sum_{j=1}^{n}\xi_{j},$$

where we let $\xi_j = \|\mathcal{T}_{\lambda}^{-\frac{1}{2}} x_j\|_H^2$ for all $j \in [n]$. Besides, it is easy to see that

$$\operatorname{tr}(\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{T}_{\mathbf{x}} - \mathcal{T})\mathcal{T}_{\lambda}^{-\frac{1}{2}}) = \frac{1}{n} \sum_{j=1}^{n} (\xi_{j} - \mathbb{E}[\xi_{j}]).$$

Using Assumption (2),

$$\xi_1 \le \frac{1}{\lambda} \|x_1\|_H^2 \le \frac{\kappa^2}{\lambda},$$

and

$$\mathbb{E}[\|\xi_1\|^2] \leq \frac{\kappa^2}{\lambda} \mathbb{E} \|\mathcal{T}_{\lambda}^{-\frac{1}{2}} x_1\|_H^2 \leq \frac{\kappa^2 \mathcal{N}(\lambda)}{\lambda}.$$

Applying Lemma 21, we get that there exists a subset V_1 of Z^n with measure at least $1 - \delta$, such that for all $z \in V_1$,

$$\operatorname{tr}(\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{T}_{\mathbf{x}} - \mathcal{T})\mathcal{T}_{\lambda}^{-\frac{1}{2}}) \leq 2\left(\frac{2\kappa^{2}}{n\lambda} + \sqrt{\frac{\kappa^{2}\mathcal{N}(\lambda)}{n\lambda}}\right)\log\frac{2}{\delta}.$$

Combining with Lemma 8, taking the union bounds, rescaling δ , and noting that

$$\begin{split} \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-1}\mathcal{T}_{\mathbf{x}}) &= \operatorname{tr}(\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\mathcal{T}_{\lambda}^{\frac{1}{2}}\mathcal{T}_{\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}\mathcal{T}_{\lambda}^{-\frac{1}{2}}\mathcal{T}_{\lambda}^{\frac{1}{2}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}) \\ &\leq & \|\mathcal{T}_{\lambda}^{\frac{1}{2}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\|^{2} \operatorname{tr}(\mathcal{T}_{\lambda}^{-\frac{1}{2}}\mathcal{T}_{\mathbf{x}}\mathcal{T}_{\lambda}^{-\frac{1}{2}}) \\ &= & \|\mathcal{T}_{\lambda}^{\frac{1}{2}}\mathcal{T}_{\mathbf{x}\lambda}^{-\frac{1}{2}}\|^{2} \left(\operatorname{tr}(\mathcal{T}_{\lambda}^{-\frac{1}{2}}(\mathcal{T}_{\mathbf{x}} - \mathcal{T})\mathcal{T}_{\lambda}^{-\frac{1}{2}}) + \mathcal{N}(\lambda)\right). \end{split}$$

we get that there exists a subset V of Z^n with measure at least $1 - \delta$, such that for all $\mathbf{z} \in V$,

$$\operatorname{tr}((\mathcal{T}_{\mathbf{x}} + \lambda)^{-1} \mathcal{T}_{\mathbf{x}}) \leq 3a_{n,\delta/2,\gamma}(\theta) \left(2\left(\frac{2\kappa^2}{n\lambda} + \sqrt{\frac{\kappa^2 \mathcal{N}(\lambda)}{n\lambda}}\right) \log \frac{4}{\delta} + \mathcal{N}(\lambda) \right),$$

which leads to the desired result using $\lambda \leq 1$, $n\lambda \geq 1$ and Assumption 3.

A.6. Proof for Corollary 5

Proof. Using a similar argument as that for (58), with $W = S_{\tilde{\mathbf{x}}}$, where $\tilde{\mathbf{x}} = \{x_1, \dots, x_m\}$, we get for any $\eta > 0$,

$$\|(I-P)\mathcal{T}^{\frac{1}{2}}\|^2 \le \eta \|(\mathcal{T}_{\tilde{\mathbf{x}}} + \eta)^{-1/2}(\mathcal{T} + \eta)^{1/2}\|^2.$$

Letting $\eta = \frac{1}{m}$, and using Lemma 8, we get that with probability at least $1 - \delta$,

$$\|(I-P)\mathcal{T}^{\frac{1}{2}}\|^2 \lesssim \frac{1}{m}\log\frac{3m^{\gamma}}{\delta}.$$

Combining with Corollary 3, one can prove the desired result.

B. Learning with Kernel Methods

Let the input space Ξ be a closed subset of Euclidean space \mathbb{R}^d , the output space $Y \subseteq \mathbb{R}$. Let μ be an unknown but fixed Borel probability measure on $\Xi \times Y$. Assume that $\{(\xi_i,y_i)\}_{i=1}^m$ are i.i.d. from the distribution μ . A reproducing kernel K is a symmetric function $K:\Xi\times\Xi\to\mathbb{R}$ such that $(K(u_i,u_j))_{i,j=1}^\ell$ is positive semidefinite for any finite set of points $\{u_i\}_{i=1}^\ell$ in Ξ . The kernel K defines a reproducing kernel Hilbert space (RKHS) $(\mathcal{H}_K,\|\cdot\|_K)$ as the completion of the linear span of the set $\{K_\xi(\cdot):=K(\xi,\cdot):\xi\in\Xi\}$ with respect to the inner product $\langle K_\xi,K_u\rangle_K:=K(\xi,u)$. For any $f\in\mathcal{H}_K$, the reproducing property holds: $f(\xi)=\langle K_\xi,f\rangle_K$.

Example B.1 (Sobolev Spaces). Let X = [0, 1] and the kernel

$$K(x, x') = \begin{cases} (1 - y)x, & x \le y; \\ (1 - x)y, & x \ge y. \end{cases}$$

Then the kernel induces a Sobolev Space $H = \{f : X \to \mathbb{R} | f \text{ is absolutely continuous }, f(0) = f(1) = 0, f \in L^2(X) \}.$

In learning with kernel methods, one considers the following minimization problem

$$\inf_{f \in \mathcal{H}_K} \int_{\Xi \times Y} (f(\xi) - y)^2 d\mu(\xi, y).$$

Since $f(\xi) = \langle K_{\xi}, f \rangle_K$ by the reproducing property, the above can be rewritten as

$$\inf_{f \in \mathcal{H}_K} \int_{\Xi \times Y} (\langle f, K_{\xi} \rangle_K - y)^2 d\mu(\xi, y).$$

Letting $X = \{K_{\xi} : \xi \in \Xi\}$ and defining another probability measure $\rho(K_{\xi}, y) = \mu(\xi, y)$, the above reduces to the learning setting in Section 2.

C. An Example for the Non-attainable Case

Let H be the usual sequence space l_2 of all infinite sequences of real numbers with its norm given by

$$\|\xi\|_2 = \left(\sum_{i=1}^{\infty} \xi_i^2\right)^{\frac{1}{2}}, \quad \forall \xi \in l_2.$$

Assume that the random vector $x=(x_1,\cdots,x_j,...,x_\infty)$ is drawn according to the distribution ρ_X , where $x_j=e_j/j$ for all j and e_1,\cdots,e_∞ are independent Bernoulli random variables. Define the regression function $f_\rho(x)=\sum_{j=1}^\infty x_j=\langle \mathbf{1},x\rangle_{l_2}$, where $\mathbf{1}=(1,\cdots,1)$. Using a direct computation, it is relatively easy to show that $f_H=f_\rho\in\overline{H_\rho}$, but $f_H\not\in H_\rho$.