

# EE378B Homework 1 Solution

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## 1 Problem 1

### 1.1 Part (a)

First we show

$$\max \{ \|\mathbf{M}\mathbf{x}\|_2 : \|\mathbf{x}\|_2 = 1 \} = \max \{ \langle \mathbf{x}, \mathbf{M}\mathbf{y} \rangle : \|\mathbf{x}\|_2 = \|\mathbf{y}\|_2 = 1 \}. \quad (1)$$

In fact the two sides are well defined since the space of  $\|\mathbf{x}\| = 1, \|\mathbf{y}\| = 1$  is compact, the supremum can be attained by a maximal point. On that space by Cauchy-Schwarz

$$\langle \mathbf{x}, \mathbf{M}\mathbf{y} \rangle \leq \|\mathbf{x}\|_2 \|\mathbf{M}\mathbf{y}\|_2 = \|\mathbf{M}\mathbf{y}\|_2, \quad (2)$$

one has LHS  $\geq$  RHS. On the other hand

$$\|\mathbf{M}\mathbf{y}\|_2 = \left\langle \frac{\mathbf{M}\mathbf{y}}{\|\mathbf{M}\mathbf{y}\|_2}, \mathbf{M}\mathbf{y} \right\rangle \leq \max \{ \langle \mathbf{x}, \mathbf{M}\mathbf{y} \rangle : \|\mathbf{x}\|_2 = \|\mathbf{y}\|_2 = 1 \}, \quad (3)$$

which implies LHS  $\leq$  RHS. Putting together completes the proof for the first part. Next we only have to show

$$\sigma_1(\mathbf{M}) = \max \{ \langle \mathbf{x}, \mathbf{M}\mathbf{y} \rangle : \|\mathbf{x}\|_2 = \|\mathbf{y}\|_2 = 1 \}. \quad (4)$$

Recall the SVD of  $\mathbf{M} \in \mathbb{R}^{m \times n}$ ,

$$\mathbf{M} = \sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^\top, \quad r \leq m \wedge n, \quad \sigma_1 \geq \sigma_2 \geq \dots \geq \sigma_r > 0, \quad (5)$$

where  $\mathbf{U} = [\mathbf{u}_1, \dots, \mathbf{u}_r] \in \mathbb{R}^{m \times r}$  and  $\mathbf{V} = [\mathbf{v}_1, \dots, \mathbf{v}_r] \in \mathbb{R}^{n \times r}$  are orthonormal. On the one hand, we have

$$\sigma_1(\mathbf{M}) = \mathbf{u}_1^\top \mathbf{M} \mathbf{v}_1 = \langle \mathbf{u}_1, \mathbf{M} \mathbf{v}_1 \rangle \leq \max \{ \langle \mathbf{x}, \mathbf{M}\mathbf{y} \rangle : \|\mathbf{x}\|_2 = \|\mathbf{y}\|_2 = 1 \}. \quad (6)$$

On the other hand, for any two normalized vectors  $\mathbf{x}$  and  $\mathbf{y}$ , we can deduce by Cauchy-Schwarz that

$$\langle \mathbf{x}, \mathbf{M}\mathbf{y} \rangle = \sum_{i=1}^r \sigma_i \langle \mathbf{x}, \mathbf{u}_i \rangle \langle \mathbf{y}, \mathbf{v}_i \rangle \leq \sigma_1 \sqrt{\sum_{i=1}^r \langle \mathbf{x}, \mathbf{u}_i \rangle^2 \sum_{i=1}^r \langle \mathbf{y}, \mathbf{v}_i \rangle^2} = \sigma_1 \|P_{\mathbf{U}} \mathbf{x}\|_2 \|P_{\mathbf{V}} \mathbf{y}\|_2 \leq \sigma_1 \|\mathbf{x}\|_2 \|\mathbf{y}\|_2 \leq \sigma_1(\mathbf{M}). \quad (7)$$

where  $P_{\mathbf{U}}, P_{\mathbf{V}}$  are projections onto subspaces  $\mathbf{U}$  and  $\mathbf{V}$ . Taken collectively, the proof is complete.

### 1.2 Part (b)

Given the SVD of  $\mathbf{M}$ , we can directly write out the SVD of  $\mathbf{M}^\top$  as

$$\mathbf{M}^\top = \sum_{i=1}^r \sigma_i \mathbf{v}_i \mathbf{u}_i^\top, \quad (8)$$

and by the result shown in part (a), we see that  $\|\mathbf{M}\|_{\text{op}} = \sigma_1(\mathbf{M}) = \sigma_1 = \sigma_1(\mathbf{M}^\top) = \|\mathbf{M}^\top\|_{\text{op}}$ . Secondly,

$$\begin{aligned}
\|\mathbf{AB}\|_{\text{op}} &= \max \{ \|\mathbf{ABx}\|_2 : \|\mathbf{x}\|_2 = 1 \} \\
&= 0 \vee \max \{ \|\mathbf{ABx}\|_2 : \|\mathbf{x}\|_2 = 1, \|\mathbf{Bx}\|_2 \neq 0 \} \\
&= 0 \vee \max \left\{ \frac{\|\mathbf{ABx}\|_2}{\|\mathbf{Bx}\|_2} \cdot \|\mathbf{Bx}\|_2 : \|\mathbf{x}\|_2 = 1, \|\mathbf{Bx}\|_2 \neq 0 \right\} \\
&\leq 0 \vee \max \left\{ \frac{\|\mathbf{AB\mathbf{y}}\|_2}{\|\mathbf{B\mathbf{y}}\|_2} \cdot \|\mathbf{B\mathbf{y}}\|_2 : \|\mathbf{x}\|_2 = 1, \|\mathbf{B\mathbf{y}}\|_2 \neq 0 \right\} \\
&\leq 0 \vee (\max \{ \|\mathbf{Az}\|_2 : \|\mathbf{z}\|_2 = 1, \mathbf{z} = \mathbf{B\mathbf{y}} \} \cdot \max \{ \|\mathbf{Bx}\|_2 : \|\mathbf{x}\|_2 = 1 \}) \\
&\leq \|\mathbf{A}\|_{\text{op}} \|\mathbf{B}\|_{\text{op}},
\end{aligned} \tag{9}$$

where the convention  $\max \emptyset = -\infty$  is used.

### 1.3 Part (c)

(i) Note that for any  $\mathbf{x}$ ,  $\|a\mathbf{Mx}\|_2 = |a|\|\mathbf{Mx}\|_2$ . By the first definition of operator norm, we get

$$\|a\mathbf{M}\|_{\text{op}} = |a|\|\mathbf{M}\|_{\text{op}}. \tag{10}$$

(ii) Again we use the first definition, and deduce that

$$\begin{aligned}
\|\mathbf{A} + \mathbf{B}\|_{\text{op}} &= \max \{ \|(\mathbf{A} + \mathbf{B})\mathbf{x}\|_2 : \|\mathbf{x}\|_2 = 1 \} \\
&= \max \{ \|\mathbf{Ax}\|_2 + \|\mathbf{Bx}\|_2 : \|\mathbf{x}\|_2 = 1 \} \\
&\leq \max \{ \|\mathbf{Ax}\|_2 + \|\mathbf{Bx}\|_2 : \|\mathbf{x}\|_2 = 1 \} \\
&\leq \max \{ \|\mathbf{Ax}\|_2 + \|\mathbf{B\mathbf{y}}\|_2 : \|\mathbf{x}\|_2 = \|\mathbf{y}\|_2 = 1 \} \\
&= \max \{ \|\mathbf{Ax}\|_2 : \|\mathbf{x}\|_2 = 1 \} + \max \{ \|\mathbf{B\mathbf{y}}\|_2 : \|\mathbf{y}\|_2 = 1 \} \\
&\leq \|\mathbf{A}\|_{\text{op}} + \|\mathbf{B}\|_{\text{op}}.
\end{aligned} \tag{11}$$

(iii) Finally, we invoke the maximum singular value definition. Since  $\|\mathbf{M}\|_{\text{op}} = \sigma_1(\mathbf{M}) = 0$ , one must have

$$\mathbf{M} = \sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^\top = \mathbf{0}. \tag{12}$$

## 2 Problem 2

### 2.1 Part (a)

Let  $\mathbf{U} = [\mathbf{u}_1, \dots, \mathbf{u}_k] \in \mathbb{R}^{n \times k}$ . For any other orthonormal basis of  $W$ , we can denote it by  $\mathbf{U}' = \mathbf{U}\mathbf{O}$ , where  $\mathbf{O} \in \mathbb{R}^{k \times k}$  is an orthogonal matrix. Then

$$\begin{aligned} \sum_{j=1}^k \langle \mathbf{u}_j, \mathbf{A}\mathbf{u}_j \rangle &= \text{tr} [\mathbf{u}_i^\top \mathbf{A}\mathbf{u}_j]_{1 \leq i, j \leq k} = \text{tr} (\mathbf{U}^\top \mathbf{A}\mathbf{U}) = \text{tr} (\mathbf{U}^\top \mathbf{A}\mathbf{U}\mathbf{O}\mathbf{O}^\top) \\ &= \text{tr} (\mathbf{O}^\top \mathbf{U}^\top \mathbf{A}\mathbf{U}\mathbf{O}) \\ &= \text{tr} (\mathbf{U}'^\top \mathbf{A}\mathbf{U}') \\ &= \sum_{j=1}^k \langle \mathbf{u}'_j, \mathbf{A}\mathbf{u}'_j \rangle, \end{aligned} \tag{13}$$

where we use the fact that  $\mathbf{O}\mathbf{O}^\top = \mathbf{I}$  and  $\text{tr}(\mathbf{A}\mathbf{B}) = \text{tr}(\mathbf{B}\mathbf{A})$  for all  $\mathbf{A} \in \mathbb{R}^{m \times n}$ ,  $\mathbf{B} \in \mathbb{R}^{n \times m}$ . Thus the definition of  $\text{tr}(\mathbf{A}|_W)$  doesn't depend on the choice of orthonormal basis.

### 2.2 Part (b)

Since

$$\dim(W \cap V_1) = 1 = \{\alpha \mathbf{u}_1 \mid \|\mathbf{u}_1\|_2 = 1, \alpha \in \mathbb{R}, \mathbf{u}_1 \in V_1\}, \tag{14}$$

we can choose a orthonormal basis of  $W$  with the first vector exactly being  $\mathbf{u}_1$ . Let the complete basis be  $\{\mathbf{u}_1, \dots, \mathbf{u}_k\}$ . Define

$$W' = \mathbf{u}_2 \oplus \dots \oplus \mathbf{u}_k, \tag{15}$$

we show that  $W' \in \mathcal{S}_{k-1}(V_2, \dots, V_k)$ . In fact since  $\mathbf{u}_1 \perp W'$ , for all  $1 \leq j \leq k$ ,

$$\begin{aligned} \dim(W' \cap V_j) &= \dim(\mathbf{u}_1 \oplus (W' \cap V_j)) - 1 \\ &= \dim((\mathbf{u}_1 \oplus W') \cap V_j) - 1 \\ &= \dim(W \cap V_j) - 1 \\ &= j - 1, \end{aligned} \tag{16}$$

where in the second line we use the fact  $\mathbf{u}_1 \in V_j$ . Next we show the desired inequality, now it's clear that

$$\begin{aligned} \text{tr}(\mathbf{A}|_W) &= \sum_{j=1}^k \langle \mathbf{u}_j, \mathbf{A}\mathbf{u}_j \rangle \\ &= \langle \mathbf{u}_1, \mathbf{A}\mathbf{u}_1 \rangle + \sum_{j=2}^k \langle \mathbf{u}_j, \mathbf{A}\mathbf{u}_j \rangle \\ &= \langle \mathbf{u}_1, \mathbf{A}\mathbf{u}_1 \rangle + \text{tr}(\mathbf{A}|_{W'}) \\ &\geq \lambda_{i_1}(\mathbf{A}) + \text{tr}(\mathbf{A}|_{W'}), \end{aligned} \tag{17}$$

while the final line follows from  $\mathbf{u}_1 \in V_1 \Rightarrow \mathbf{u}_1 = \sum_{l=1}^{i_1} \langle \mathbf{u}_1, \mathbf{v}_l \rangle \mathbf{v}_l$ , and thus

$$\langle \mathbf{u}_1, \mathbf{A}\mathbf{u}_1 \rangle = \sum_{l=1}^{i_1} \langle \mathbf{u}_1, \mathbf{v}_l \rangle^2 \lambda_l(\mathbf{A}) \geq \lambda_{i_1}(\mathbf{A}) \sum_{l=1}^{i_1} \langle \mathbf{u}_1, \mathbf{v}_l \rangle^2 = \lambda_{i_1}(\mathbf{A}). \tag{18}$$

### 2.3 Part (c)

Let  $W_0 := W$ , we prove by induction that there exists  $W_j \in \mathcal{S}_{k-j}(V_{j+1}, \dots, V_k)$  (in particular,  $W_k = \emptyset$ ) such that

$$\mathrm{tr}(\mathbf{A}|_{W_0}) \geq \lambda_{i_1}(\mathbf{A}) + \dots + \lambda_{i_j}(\mathbf{A}) + \mathrm{tr}(\mathbf{A}|_{W_j}), \quad \forall 1 \leq j \leq k. \quad (19)$$

By part (b) we see that the induction hypothesis holds for  $j = 1$ . Suppose it's true for some  $j < k$ . We invoke part (b) for  $W_j \in \mathcal{S}_{k-j}(V_{j+1}, \dots, V_k)$ . Then there must be some  $W_{j+1} \in \mathcal{S}_{k-j-1}(V_{j+2}, \dots, V_k)$  such that

$$\mathrm{tr}(\mathbf{A}|_{W_j}) \geq \lambda_{i_{j+1}}(\mathbf{A}) + \mathrm{tr}(\mathbf{A}|_{W_{j+1}}). \quad (20)$$

The conclusion also holds for  $j + 1$ . The proof is complete by induction and using  $W_k = \emptyset$ . Finally by definition, we thus have for this special choice of  $(V_1, \dots, V_k)$ ,

$$\mathcal{R}(\mathbf{A}; i_1, \dots, i_k) \geq \inf_{W \in \mathcal{S}_k(V_1, \dots, V_k)} \mathrm{tr}(\mathbf{A}|_W) \geq \lambda_{i_1}(\mathbf{A}) + \dots + \lambda_{i_k}(\mathbf{A}). \quad (21)$$

### 2.4 Part (d)

We use the same notations  $\mathbf{v}_1, \dots, \mathbf{v}_n$  in part (b) for  $\mathbf{A}$ 's eigenvectors. We construct  $W$  in the following way. Let  $U_1 := \mathrm{span}(\mathbf{v}_{i_1}, \dots, \mathbf{v}_n)$ , then  $\dim(U_1) = n - i_1 + 1$ , while  $\dim(V_1) = i_1$ , thus

$$\dim(V_1 \cap U_1) \geq 1, \quad (22)$$

we can choose some  $\mathbf{u}_1 \in V_1 \cap U_1$  of unit norm. Then we can simply choose normalized vectors  $\mathbf{u}_j \in V_j \setminus V_{j-1}$ ,  $\mathbf{u}_j \perp V_{j-1}$  for  $j = 2, \dots, k$  by Gram-Schmidt orthogonalization since  $V_j \setminus V_{j-1} \neq \emptyset$ . Let  $W = \mathrm{span}(\mathbf{u}_1, \dots, \mathbf{u}_k)$  and  $W' = \mathrm{span}(\mathbf{u}_2, \dots, \mathbf{u}_k)$ . Clearly since  $\mathbf{u}_1, \dots, \mathbf{u}_k$  are orthonormal, and for any  $1 \leq j \leq k$ ,  $\mathbf{u}_1, \dots, \mathbf{u}_j \in V_j$ ,  $\mathbf{u}_{j+1}, \dots, \mathbf{u}_k \perp V_j$ , one has

$$\dim(W \cap V_j) = \dim(\mathrm{span}(\mathbf{u}_1, \dots, \mathbf{u}_j)) = j. \quad (23)$$

Following the same argument, one can easily check  $\dim(W' \cap V_j) = j - 1$  for all  $1 \leq j \leq k$ . Therefore

$$\mathrm{tr}(\mathbf{A}|_W) = \langle \mathbf{u}_1, \mathbf{A}\mathbf{u}_1 \rangle + \mathrm{tr}(\mathbf{A}|_{W'}) \leq \lambda_{i_1}(\mathbf{A}) + \mathrm{tr}(\mathbf{A}|_{W'}), \quad (24)$$

where the inequality follows from  $\mathbf{u}_1 \in U_1$ , and

$$\langle \mathbf{u}_1, \mathbf{A}\mathbf{u}_1 \rangle = \sum_{l=i_1}^n \langle \mathbf{u}_1, \mathbf{v}_l \rangle^2 \lambda_l(\mathbf{A}) \leq \lambda_{i_1}(\mathbf{A}) \sum_{l=i_1}^n \langle \mathbf{u}_1, \mathbf{v}_l \rangle^2 = \lambda_{i_1}(\mathbf{A}). \quad (25)$$

### 2.5 Part (e) - Optional

We can construct the chain of  $W =: W_0 \supset W_1 \cdots \supset W_k = \emptyset$  from below by induction. First of all, we want to construct a perturbed orthonormal basis of  $\mathbb{R}^n$ , such that

$$[\mathbf{g}_1, \dots, \mathbf{g}_n] =: \mathbf{G} = \mathbf{V}\mathbf{O}, \quad \|\mathbf{O} - \mathbf{I}\|_{\max} \leq \epsilon,$$

where  $\mathbf{V}$  is the eigenvector matrix of  $\mathbf{A}$  and  $\mathbf{O}$  is orthogonal. We define  $U_j := \mathrm{span}(\mathbf{g}_{i_j}, \dots, \mathbf{g}_n)$  for  $1 \leq j \leq k$ . We claim there exists a  $\mathbf{G}$  for any  $\epsilon > 0$  such that

$$U_j \cap V_{j-1} = \emptyset, \quad \forall j = 2, 3, \dots, k. \quad (26)$$

**Lemma 1.** For  $1 \leq j \leq k$  and any  $W_j \in \mathcal{S}_{k-j}(V_{j+1}, \dots, V_k)$  that takes the form  $W_j = \mathrm{span}(\mathbf{u}_{j+1}, \dots, \mathbf{u}_n)$  where the  $n$  orthonormal basis  $\mathbf{u}_{j+1}, \dots, \mathbf{u}_k$  satisfy  $\mathbf{u}_l \in U_l$  for  $j + 1 \leq l \leq k$ , there exists a  $W_{j-1} \in \mathcal{S}_{k-j+1}(V_j, \dots, V_k)$  such that  $W_{j-1} = \mathrm{span}(\mathbf{u}_j, \dots, \mathbf{u}_n) = \mathrm{span}(\mathbf{u}_j, W_j)$  where  $\mathbf{u}_j \in U_j$ .

*Proof.* Since  $\dim(U_j) = n - i_j + 1$  and  $\dim(V_1) = i_j$ , thus

$$\dim(V_j \cap U_j) \geq 1 \quad (27)$$

and we can choose some  $\tilde{\mathbf{u}}_j \in V_j \cap U_j$  of unit norm. For any  $j+1 \leq l \leq k$ , since  $\mathbf{u}_l \in W_j$  and  $W_j \cap V_j \subset U_{j+1} \cap V_j = \emptyset$ , we know  $\tilde{\mathbf{u}}_j$  is linearly independent of  $\mathbf{u}_{j+1}, \dots, \mathbf{u}_k$ . Therefore, by setting  $W_{j-1} = \text{span}(\tilde{\mathbf{u}}_j, W_j)$ , we have

$$\dim(W_{j-1}) = \dim(W_j) + 1 = j, \quad (28)$$

$$\dim(W_{j-1} \cap V_l) = \dim((W_j + \tilde{\mathbf{u}}_j) \cap V_l) = \dim(W_j \cap V_l) + 1 = j, \quad \forall j \leq l \leq k, \quad (29)$$

where the second line comes from  $\tilde{\mathbf{u}}_j \notin W_j$  and  $\tilde{\mathbf{u}}_j \in V_j \subset V_l$ . Therefore  $W_{j-1} \in \mathcal{S}_{k-j+1}(V_j, \dots, V_k)$ . Finally, we take  $\mathbf{u}_j$  by Gram-Schmidt orthogonalization

$$\mathbf{u}_j = \frac{\tilde{\mathbf{u}}_j - \sum_{l=j+1}^n \langle \tilde{\mathbf{u}}_j, \mathbf{u}_l \rangle \mathbf{u}_l}{\left\| \tilde{\mathbf{u}}_j - \sum_{l=j+1}^n \langle \tilde{\mathbf{u}}_j, \mathbf{u}_l \rangle \mathbf{u}_l \right\|_2}. \quad (30)$$

Note that  $\tilde{\mathbf{u}}_j \in U_j, \mathbf{u}_l \in U_l \subset U_j$  for all  $j+1 \leq l \leq n$ , one must have  $\mathbf{u}_j \in U_j$ . The proof of the lemma is complete.  $\square$

Therefore, by iteratively invoking the lemma starting from  $W_k := \emptyset$ , we can construct a sequence of subspace  $W_j \in \mathcal{S}_{k-j}(V_{j+1}, \dots, V_k)$  where  $W := W_0$  has an orthonormal basis  $\mathbf{u}_1, \dots, \mathbf{u}_k$  such that  $\mathbf{u}_j \in U_j$ . Note that

$$\langle \mathbf{u}_j, \mathbf{A} \mathbf{u}_j \rangle \leq \max_{\|\mathbf{x}\|_2=1, \mathbf{x} \in U_j} \langle \mathbf{x}, \mathbf{A} \mathbf{x} \rangle. \quad (31)$$

Clearly for any  $\delta > 0$ , we can always choose  $\epsilon > 0$  small enough such that

$$\langle \mathbf{u}_j, \mathbf{A} \mathbf{u}_j \rangle \leq \max_{\|\mathbf{x}\|_2=1, \mathbf{x} \in \text{span}(\mathbf{v}_{i_j}, \dots, \mathbf{v}_n)} \langle \mathbf{x}, \mathbf{A} \mathbf{x} \rangle + \delta \leq \lambda_{i_j}(\mathbf{A}) + \delta, \quad \forall 1 \leq j \leq k, \quad (32)$$

since

$$\lim_{\epsilon \rightarrow 0} [\mathbf{g}_{i_j}, \dots, \mathbf{g}_n] \rightarrow [\mathbf{v}_{i_j}, \dots, \mathbf{v}_n], \quad \forall 1 \leq j \leq k. \quad (33)$$

Therefore

$$\text{tr}(\mathbf{A}|_W) = \sum_{j=1}^k \langle \mathbf{u}_j, \mathbf{A} \mathbf{u}_j \rangle \leq \sum_{j=1}^k \lambda_{i_j}(\mathbf{A}) + k\delta. \quad (34)$$

Note that  $\delta$  can be arbitrarily small, we thus can conclude

$$\inf_{W \in \mathcal{S}_k(V_1, \dots, V_k)} \text{tr}(\mathbf{A}|_W) \leq \sum_{j=1}^k \lambda_{i_j}(\mathbf{A}). \quad (35)$$

We only need to prove for any  $\epsilon > 0$ ,  $\mathbf{G}$  exists such that

$$U_j \cap V_{j-1} = \emptyset, \quad \forall j = 2, 3, \dots, k. \quad (36)$$

In fact for any  $2 \leq j \leq k$ ,  $\dim(V_{j-1}) \leq n - 1$ . Under the basis  $\mathbf{v}_1, \dots, \mathbf{v}_n$ ,  $V_{j-1}$  is a lower dimensional hyperplane inside  $\mathbb{R}^n$ . While  $U_j$  has an induced Haar measure from the orthogonal transform  $\mathbf{O}$ . The Haar measure is uniform and therefore the measure of its intersection with a fixed lower dimensional hyperplane being nonempty is 0. Hence such  $\mathbf{G}$  must exist since  $\|\mathbf{O} - \mathbf{I}\|_{\max} \leq \epsilon$  has positive Haar measure.

### 3 Problem 3

We first show for any subspace  $W$  of dimension  $k$ ,

$$\text{tr}(\mathbf{B}|_W) \leq \lambda_1(\mathbf{B}) + \cdots + \lambda_k(\mathbf{B}). \quad (37)$$

Suppose the  $i$ -th normalized eigenvector of  $\mathbf{B}$  is  $\mathbf{z}_i$ , let  $\mathbf{U} = [\mathbf{u}_1, \dots, \mathbf{u}_k] \in \mathbb{R}^{n \times k}$  be an orthonormal basis of  $W$ . Let  $\mathbf{Z} = [\mathbf{z}_1, \dots, \mathbf{z}_n] \in \mathbb{R}^{n \times n}$ . Note that

$$\mathbf{B} = \sum_{i=1}^n \lambda_i(\mathbf{B}) \mathbf{z}_i \mathbf{z}_i^\top. \quad (38)$$

Then

$$\text{tr}(\mathbf{B}|_W) = \sum_{j=1}^k \langle \mathbf{u}_j, \mathbf{B} \mathbf{u}_j \rangle = \sum_{j=1}^k \sum_{i=1}^n \lambda_i(\mathbf{B}) \langle \mathbf{u}_j, \mathbf{z}_i \rangle^2 := \sum_{i=1}^n \lambda_i(\mathbf{B}) \sum_{j=1}^k c_{ij}, \quad (39)$$

where  $c_{ij} = \langle \mathbf{z}_i, \mathbf{u}_j \rangle^2$  and

$$0 \leq p_i := \sum_{j=1}^k c_{ij} \leq 1, \quad \sum_{i=1}^n p_i = \sum_{j=1}^k \sum_{i=1}^n c_{ij} = \sum_{j=1}^k 1 = k. \quad (40)$$

Thus given  $\lambda_1(\mathbf{B}) \geq \cdots \geq \lambda_n(\mathbf{B})$  and at most  $k$  steps of greedy algorithm, we can show

$$\text{tr}(\mathbf{B}|_W) = \sum_{i=1}^n \lambda_i(\mathbf{B}) p_i \leq \sum_{i=1}^n \lambda_i(\mathbf{B}) \mathbf{1}_{i \leq k} = \lambda_1(\mathbf{B}) + \cdots + \lambda_k(\mathbf{B}). \quad (41)$$

Similarly

$$\text{tr}(\mathbf{B}|_W) = \sum_{i=1}^n \lambda_i(\mathbf{B}) p_i \geq \sum_{i=1}^n \lambda_i(\mathbf{B}) \mathbf{1}_{i \geq n-k+1} = \lambda_{n-k+1}(\mathbf{B}) + \cdots + \lambda_n(\mathbf{B}). \quad (42)$$

Finally we can finish the proof using Problem 2, which tells us

$$\begin{aligned} \lambda_{i_1}(\mathbf{A} + \mathbf{B}) + \cdots + \lambda_{i_k}(\mathbf{A} + \mathbf{B}) &= \mathcal{R}(\mathbf{A} + \mathbf{B}; i_1, \dots, i_k) \\ &= \sup_{(V_1, \dots, V_k) \in \mathcal{F}(i_1, \dots, i_k)} \inf_{W \in \mathcal{S}_k(V_1, \dots, V_k)} \text{tr}(\mathbf{A} + \mathbf{B}|_W) \\ &= \sup_{(V_1, \dots, V_k) \in \mathcal{F}(i_1, \dots, i_k)} \inf_{W \in \mathcal{S}_k(V_1, \dots, V_k)} (\text{tr}(\mathbf{A}|_W) + \text{tr}(\mathbf{B}|_W)) \\ &\leq \sup_{(V_1, \dots, V_k) \in \mathcal{F}(i_1, \dots, i_k)} \inf_{W \in \mathcal{S}_k(V_1, \dots, V_k)} (\text{tr}(\mathbf{A}|_W) + \lambda_1(\mathbf{B}) + \cdots + \lambda_k(\mathbf{B})) \\ &= \left( \sup_{(V_1, \dots, V_k) \in \mathcal{F}(i_1, \dots, i_k)} \inf_{W \in \mathcal{S}_k(V_1, \dots, V_k)} \text{tr}(\mathbf{A}|_W) \right) + \lambda_1(\mathbf{B}) + \cdots + \lambda_k(\mathbf{B}) \\ &= \mathcal{R}(\mathbf{A}; i_1, \dots, i_k) + \lambda_1(\mathbf{B}) + \cdots + \lambda_k(\mathbf{B}) \\ &= \lambda_{i_1}(\mathbf{A}) + \cdots + \lambda_{i_k}(\mathbf{A}) + \lambda_1(\mathbf{B}) + \cdots + \lambda_k(\mathbf{B}). \end{aligned} \quad (43)$$

Similarly, we can get the inequality in the other direction,

$$\lambda_{i_1}(\mathbf{A} + \mathbf{B}) + \cdots + \lambda_{i_k}(\mathbf{A} + \mathbf{B}) \geq \lambda_{i_1}(\mathbf{A}) + \cdots + \lambda_{i_k}(\mathbf{A}) + \lambda_{n-k+1}(\mathbf{B}) + \cdots + \lambda_n(\mathbf{B}). \quad (44)$$