# Stat461 - 2023 Fall HW2

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## Exercise 1

1.(a)

 $\Rightarrow$  We prove by contradiction. Suppose T is unbounded, we can find a sequence  $t_n \in T$  such that  $||t_n|| \to \infty$ . Then we consider the set  $T \cup (-T) = \{t1\} \cup \{-t\}$ , using the symmetry of  $g \sim \mathcal{N}(0, I)$ , we have

$$w(T) = w(-T) \geq \frac{1}{2} w(T \cup (-T))$$

and we have  $T_n := \{t_n, -t_n\} \in T \cup (-T)$ . For such  $T_n$  we notice that

$$w(T_n) = \mathbb{E}\left[\sup_{t \in \{t_n, -t_n\}} \langle t, g \rangle\right]$$
$$= \mathbb{E}\left[\left|\mathcal{N}(0, ||t_n||^2)\right|\right]$$
$$\geq ||t_n|| \mathbb{E}\left[\left|\mathcal{N}(0, 1)\right|\right]$$
$$= ||t_n|| \sqrt{2/\pi}$$

so using  $T_n \subset T \cup (-T)$ 

$$w(T) \ge \frac{1}{2}w(T \cup (-T)) \ge \frac{1}{2}w(T_n) \ge \frac{1}{2} ||t_n|| \sqrt{2/\pi} \to \infty$$

thus if T is unbounded, w(T) is unbounded. Here we have the contradiction so if w(T) is finite, T is bounded.

 $\Leftarrow$  If T is bounded, say  $||t|| \le \tau$ ,  $\forall t \in T$ , we have

$$w(T) = \mathbb{E} \left[ \sup_{t \in T} \langle t, g \rangle \right]$$

$$\leq \mathbb{E} \left[ \sup_{t \in T} ||t|| \cdot ||g|| \right]$$

$$\leq \mathbb{E} \left[ \tau ||g|| \right]$$

$$\leq \tau \mathbb{E} \left[ \sqrt{\chi_n^2} \right]$$

$$\leq \tau \sqrt{\mathbb{E} \left[ \chi_n^2 \right]}$$

$$= \tau \sqrt{n} < \infty$$

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1.(b)

Using the unitarity of  $g \sim Ug \sim \mathcal{N}(0, I)$ , we have

$$w(UT + y) = \mathbb{E} \left[ \sup_{t \in T} \langle Ut + y, g \rangle \right]$$

$$= \mathbb{E} \left[ \sup_{t \in T} \langle U'(Ut + y), g \rangle \right]$$

$$= \mathbb{E} \left[ \sup_{t \in T} \langle t + Uy, g \rangle \right]$$

$$= \mathbb{E} \left[ \sup_{t \in T} \langle t, g \rangle + \langle Uy, g \rangle \right]$$

$$= \mathbb{E} \left[ \sup_{t \in T} \langle t, g \rangle \right] + \mathbb{E} \left[ \langle Uy, g \rangle \right]$$

$$= w(T) + \langle Uy, 0 \rangle$$

$$= w(T)$$

1.(c)

For any  $t_1, t_2 \in T$  and  $\alpha \in \mathbb{R}$ , we have  $t_\alpha := \alpha t_1 + (1 - \alpha)t_2 \in \text{conv}(T)$ . And we have for any  $g \in \mathbb{R}$ 

$$\langle t_{\alpha}, g \rangle = \alpha \langle t_{1}, g \rangle + (1 - \alpha) \langle t_{2}, g \rangle \leq \sup_{t \in \{t_{1}, t_{2}\}} \langle t, g \rangle$$

the above statement holds for any  $t_1, t_2 \in T$ , thus prove the statement (because for any  $\tilde{t} \in \text{conv}(T)$  we can always find such  $t_1, t_2 \in T$  and  $\alpha$  that  $\tilde{t} = \alpha t_1 + (1 - \alpha)t_2$ ).

1.(d)

• We have

$$w(T+S) = \mathbb{E}\left[\sup_{r \in T+S} \langle r, g \rangle\right]$$

$$= \mathbb{E}\left[\sup_{t \in T, s \in S} \langle t+s, g \rangle\right]$$

$$= \mathbb{E}\left[\sup_{t \in T} \langle t, g \rangle + \sup_{s \in S} \langle s, g \rangle\right]$$

$$= w(T) + w(S)$$

• We have

$$w(aT) = \mathbb{E}\left[\sup_{r \in aT} \langle r, g \rangle\right]$$

$$= \mathbb{E}\left[\sup_{t \in T} \langle at, g \rangle\right]$$

$$= \begin{cases} a\mathbb{E}\left[\sup_{t \in T} \langle t, g \rangle\right] & a \ge 0 \\ -a\mathbb{E}\left[\sup_{t \in T} \langle t, g \rangle\right] & a < 0 \end{cases}$$

$$= |a| w(T)$$

1.(e)

 $w(T) = \frac{1}{2}w(T) + \frac{1}{2}w(-T) = \frac{1}{2}w(T-T)$  is trivial using the previous results. Thus we have

$$w(T) = \frac{1}{2}w(T - T) = \frac{1}{2}\mathbb{E}\left[\sup_{x,y \in T} \langle x - y, g \rangle\right]$$

1.(f)

It suffices the consider a centered (in the sense that  $||t|| \le b = \text{diam}(T)/2$ ) set T due to previous arguments that w(T) is invariant under unitary and translation. We have

(i) Already proved in (a). Assume  $\tilde{t}$  being the 'boundary' point of T, i.e.  $\|\tilde{t}\| = b$ , we have

$$w(T) \ge \frac{1}{2} w(T \cup (-T)) \ge \frac{1}{2} w(\{\tilde{t}, -\tilde{t}\}) \ge \frac{1}{2} \|\tilde{t}\| \sqrt{2/\pi} = b\sqrt{2/\pi} = \frac{\operatorname{diam}(T)}{\sqrt{2\pi}}$$

(ii) Already proved in (b).

$$w(T) = \mathbb{E} \left[ \sup_{t \in T} \langle t, g \rangle \right]$$

$$\leq \mathbb{E} \left[ \sup_{t \in T} ||t|| \cdot ||g|| \right]$$

$$\leq \mathbb{E} \left[ b ||g|| \right]$$

$$\leq b \mathbb{E} \left[ \sqrt{\chi_n^2} \right]$$

$$\leq b \sqrt{\mathbb{E} \left[ \chi_n^2 \right]}$$

$$= \frac{\sqrt{n}}{2} \operatorname{diam}(T)$$

Together we have

$$\frac{\operatorname{diam}(T)}{\sqrt{2\pi}} \le w(T) \le \frac{\sqrt{n}}{2} \operatorname{diam}(T)$$

### Exercise 2

Proof for

$$\sqrt{\frac{2}{\pi}\sum_{i=1}^{n}a_{i}^{2}}\overset{(i)}{\leq}\sqrt{\frac{2}{\pi}}\mathcal{R}(T)\overset{(ii)}{\leq}\mathcal{G}(T)\overset{(iii)}{\leq}\sqrt{\sum_{i=1}^{n}a_{i}^{2}}$$

(i) Note that T is symmetric w.r.t. the origin, so the sup is always achieved at  $sgn(t_i) = \varepsilon_i$ :

$$\mathcal{R}(T) = \mathbb{E}\left[\sup_{t:\sum t_i^2/a_i^2 \le 1} \sum_{i=1}^n \varepsilon_i t_i\right]$$

$$= \sup_{t:\sum t_i^2/a_i^2 \le 1} ||t||_1$$

$$= \sup_{\tilde{t} \in \mathbb{S}^{n-1}} \sum_{i=1}^n |a_i \tilde{t}_i|$$

The supermum is achieved at  $\tilde{t} \propto (a_1, \dots, a_n)$ , so we have  $\mathcal{R}(T) = \sqrt{\sum_{i=1}^n a_i^2}$ .

(ii) We have for any T:

$$\mathcal{G}(T) = \mathbb{E}_g \left[ \sup_{t \in T} \sum_{i=1}^n g_i t_i \right]$$

$$= \mathbb{E}_{g,\varepsilon} \left[ \sup_{t \in T} \sum_{i=1}^n \varepsilon_i |g_i| t_i \right]$$

$$\geq \mathbb{E}_{\varepsilon} \left[ \sup_{t \in T} \sum_{i=1}^n \varepsilon_i \mathbb{E}_g [|g_i|] t_i \right]$$

$$= \sqrt{\frac{2}{\pi}} \mathcal{R}(T)$$

(iii) We have

$$\mathcal{G}(T) = \mathbb{E}_g \left[ \sup_{t: \sum t_i^2 / a_i^2 \le 1} \sum_{i=1}^n g_i t_i \right]$$
$$= \mathbb{E}_g \left[ \sup_{\tilde{t} \in \mathbb{S}^{n-1}} \sum_{i=1}^n a_i g_i \tilde{t}_i \right]$$

similarly, the supermum is achieved at  $\tilde{t} \propto (a_1 g_i, \dots, a_n g_n)$ , then

$$\mathcal{G}(T) = \mathbb{E}_g \left[ \sqrt{\sum_{i=1}^n a_i^2 g_i^2} \right]$$

$$\leq \sqrt{\mathbb{E}_g \left[ \sum_{i=1}^n a_i^2 g_i^2 \right]}$$

$$= \sqrt{\sum_{i=1}^n a_i^2}$$

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### Exercise 3

1. Construct the covering set sequence  $\{\mathcal{N}(T,\varepsilon_i)\}_{i=k}^K$ , with  $\varepsilon_i$  and k,K chosen as follows:

$$\begin{split} \varepsilon_i = & 2^{-i} \\ k : & 2^{-k} \le \operatorname{diam}(T) \le 2^{-k+1} \\ K : & 2^{-K-1} \le \frac{\kappa w(T)}{\sqrt{n}} \le 2^{-K} \end{split}$$

for some pre-determined small  $\kappa$ . i.e. we have  $\operatorname{diam}(T) \sim \varepsilon_k \stackrel{\varepsilon_i = 2^{-i}}{\searrow} \varepsilon_K \sim \frac{\kappa w(T)}{\sqrt{n}}$ .

Using the covering sequence, for each given  $t \in T$ , we can define maps  $\pi_i(t)$  as:

$$\pi_i(t) = t_i \in \mathcal{N}(T, \varepsilon_i), \ s.t. \ \pi_i(t) \in \mathcal{N}(t, \varepsilon_{i-1})$$

Then we have

$$w(T) = \mathbb{E}\left[\sup_{t \in T} \langle t, g \rangle\right]$$

$$\leq \sum_{i=k}^{K} \mathbb{E}\left[\sup_{t \in T} \langle \pi_{i}(t) - \pi_{i-1}(t), g \rangle\right] + \mathbb{E}\left[\sup_{t \in T} \langle t - \pi_{K}(t), g \rangle\right]$$

2. For  $t - \pi_K(t)$  term, since K satisfy  $a \leq \frac{w(T)}{\sqrt{n}}$  covering, we have

$$\mathbb{E}\left[\sup_{t\in T} \langle t - \pi_K(t), g \rangle\right] \le \|t - \pi_K(t)\| \,\mathbb{E}\left[\|g\|\right] \le \varepsilon_K \sqrt{n} \le \kappa w(T)$$

3. For  $\pi_i(t) - \pi_{i-1}(t)$ , which is  $\varepsilon_{i-1}$  bounded and there are at most  $N_2(T, \varepsilon_{i-1})$  possible choices of  $\pi_i(t) - \pi_{i-1}(t)$ , we can use maximal inequality for sub-Gaussian random variables to get

$$\mathbb{E}\left[\sup_{t\in T} \langle \pi_i(t) - \pi_{i-1}(t), g \rangle\right] \lesssim \varepsilon_{i-1} \sqrt{N_2(T, \varepsilon_{i-1})} \leq s(T)$$

and there are  $\sim (K - k)$  terms in the sum, so we have

$$w(T) \le C(K - k)s(T) + \mathbb{E}\left[\sup_{t \in T} \langle t - \pi_k(t), g \rangle\right]$$
$$\le C(K - k)s(T) + \kappa w(T)$$
$$\Rightarrow w(T) \le \frac{C(K - k)}{1 - \kappa}s(T) \lesssim ks(T)$$

4. Now we analyze the scale of K-k: Note that we chose  $2^{-k} \sim \operatorname{diam}(T)$  and  $2^{-K} \sim \frac{\kappa w(T)}{\sqrt{n}}$ . Thus

$$K - k = -\log_2 \frac{\kappa w(T)}{\sqrt{n} \operatorname{diam}(T)} \le \log_2 \frac{\sqrt{2\pi n}}{\kappa} \lesssim \log_2 n$$

To summarize, we have

$$w(T) \lesssim s(T) \log(n)$$

#### Exercise 4 MJW 8.3

We first show Courant-Fischer variational representation of eigenvalue, given in Exercise 8.1. We have

$$\min_{\mathbb{V}\in\mathcal{V}_{j-1}} \max_{x\in\mathbb{S}^{n-1}\cap\mathbb{V}^{\perp}} \langle Qx, x\rangle = \min_{\mathbb{V}\in\mathcal{V}_{j-1}} \max_{\nu} \left\langle \sum_{i=j}^{n} \nu_i v_i, Q \sum_{i=j}^{n} \nu_i v_i \right\rangle$$

in which  $\{v_i\}$  is an orthonormal basis of  $\mathbb{V}$ . The transformation between  $\{v_i\}$  and  $\{q_i\}$  is denoted V = QP, then

$$\begin{split} \min_{\mathbb{V}\in\mathcal{V}_{j-1}} \max_{\nu} \left\langle \sum_{i=j}^{n} \nu_{i} v_{i}, Q \sum_{i=j}^{n} \nu_{i} v_{i} \right\rangle &= \min_{\mathbb{V}\in\mathcal{V}_{j-1}} \max_{\nu} \left\langle \sum_{i=j}^{n} \nu_{i} \sum_{k=1}^{n} P_{ki} q_{k}, Q \sum_{i=j}^{n} \nu_{i} \sum_{k=1}^{n} P_{ki} q_{k} \right\rangle \\ &= \min_{\mathbb{V}\in\mathcal{V}_{j-1}} \max_{\nu} \sum_{i,\tilde{i}} \sum_{k,\tilde{k}} \nu_{i} \nu_{\tilde{i}} P_{ki} P_{\tilde{k}\tilde{i}} \left\langle q_{k}, Q q_{\tilde{k}} \right\rangle \\ &= \min_{\mathbb{V}\in\mathcal{V}_{j-1}} \max_{\nu} \sum_{i,\tilde{i}} \sum_{k,\tilde{k}} \nu_{i} \nu_{\tilde{i}} P_{ki} P_{\tilde{k}\tilde{i}} \delta_{k,\tilde{k}} \gamma_{k} \\ &= \min_{\mathbb{V}\in\mathcal{V}_{j-1}} \max_{\nu} \sum_{i,\tilde{i}} \sum_{k} \nu_{i} \nu_{\tilde{i}} P_{ki} P_{k\tilde{i}} \gamma_{k} \\ &= \min_{P\in SO(n)} \max_{\|\nu\|=1,\nu\in\mathbb{R}^{n-j+1}} \sum_{k} \gamma_{k} (\sum_{i=j}^{n} P_{ki} \nu_{i})^{2} \end{split}$$

note that we have

$$\sum_{k=1}^{n} \left(\sum_{i=j}^{n} P_{ki} \nu_{i}\right)^{2} = \sum_{i,j} \nu_{i} \nu_{j} \sum_{k=1}^{n} P_{ki} P_{kj} = \sum_{i,j} \nu_{i} \nu_{j} \delta_{ij} = \|\nu\|^{2} = 1$$

thus the above equation has max reached when

$$\operatorname*{arg\,min}_{k} \sum_{i=j}^{n} P_{ki} \nu_{i} = 1$$

and then the min is reached when the first j-1 rows of P is of shape

$$\begin{bmatrix} \tilde{P}_{(j-1)\times(j-1)} & 0 \end{bmatrix}$$

and the extreme value is  $\gamma_i$ .

Using the representation we have  $\forall i, j \in [n]$ :

$$\gamma_i(A) = \min_{\mathbb{V} \in \mathcal{V}_{i-1}} \max_{x \in \mathbb{S}^{n-1} \cap \mathbb{V}^{\perp}} \langle Ax, x \rangle$$

$$\gamma_j(B) = \min_{\mathbb{V} \in \mathcal{V}_{j-1}} \max_{x \in \mathbb{S}^{n-1} \cap \mathbb{V}^{\perp}} \langle Bx, x \rangle$$

denote the corresponding  $\mathbb{V}$  as  $\mathbb{V}_A$  and  $\mathbb{V}_B$ . The vector being  $x_A$  and  $x_B$  respectively. Note that we have  $\dim(\mathbb{V}_A \cup \mathbb{V}_B) \leq i + j - 2$ , i.e.  $\mathcal{V}_{i-1} \cup \mathcal{V}_{j-1} \subset \mathcal{V}_{i+j-2}$ , we would have

$$\begin{split} \gamma_{i+j-1}(A+B) &= \min_{\mathbb{V} \in \mathcal{V}_{i+j-2}} \max_{x \in \mathbb{S}^{n-1} \cap \mathbb{V}^{\perp}} \left\langle (A+B)x, x \right\rangle \\ &\leq \min_{\mathbb{V} \in \mathcal{V}_{i+j-2}} \max_{x \in \mathbb{S}^{n-1} \cap \mathbb{V}^{\perp}} \left\langle Ax, x \right\rangle + \max_{x \in \mathbb{S}^{n-1} \cap \mathbb{V}^{\perp}} \left\langle Bx, x \right\rangle \\ &\leq \min_{\mathbb{V} \in \mathcal{V}_{i+j-2}} \left\langle Ax_A, x_A \right\rangle + \left\langle Bx_B, x_B \right\rangle \\ &\leq \gamma_i(A) + \gamma_j(B) \end{split}$$

Taking j = 1, A = Q, B = R we have

$$\gamma_i(Q) - \gamma_i(R) \le \gamma_1(Q - R) = |||Q - R|||_2$$