

Rules: Discussion of the problems is permitted, but writing the assignment together is not (i.e. you are not allowed to see the actual pages of another student).

This homework has 200 points, and its weight is the same as those of Homeworks 1 and 2 together.

1. (15 points) **Some Technical Proofs from Johnson-Lindenstrauss Lemma**

(a) Suppose  $g$  is a random variable with normal distribution  $N(0, 1)$ . Prove the following.

- i. For odd  $n \geq 1$ ,  $E[g^n] = 0$ .
- ii. For even  $n \geq 2$ ,  $E[g^n] \geq 1$ .

(Hint: Use induction. Let  $I_n := E[g^n] = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} x^n e^{-\frac{x^2}{2}} dx$ . Use integration by parts to show that  $I_{n+2} = (n+1)I_n$ .)

(b) Suppose  $\gamma_j$ 's are independent uniform  $\{-1, 1\}$ -random variables and  $g_j$ 's are independent random variables, each having normal distribution  $N(0, 1)$ . Suppose  $v_j$ 's are real numbers, and define  $X := (\sum_j \gamma_j v_j)^2$  and  $\hat{X} := (\sum_j g_j v_j)^2$ . Show that for all integers  $n \geq 1$ ,  $E[X^n] \leq E[\hat{X}^n]$ .

(c) Suppose  $Z$  is a random variable having normal distribution  $N(0, \nu^2)$ . Compute  $E[e^{tZ^2}]$ . For what values of  $t$  is your expression valid?

2. (25 points) **Can Johnson-Lindenstrauss Lemma preserve area?**

(a) Suppose the distances between three points are preserved with multiplicative error  $\epsilon$ . Is the area of the corresponding triangle also always preserved with multiplicative error  $O(\epsilon)$ , or even some constant multiplicative error?

(b) Suppose  $u$  and  $v$  are mutually orthogonal unit vectors. Observe that the vectors  $u$  and  $v$  together with the origin form a right-angled isosceles triangle with area  $\frac{1}{2}$ . Suppose the lengths of the triangle are distorted with multiplicative error at most  $\epsilon$ . What is the multiplicative error for the area of the triangle?

(c) Suppose a set  $V$  of  $n$  points are given in Euclidean space  $\mathbb{R}^n$ . Let  $0 < \epsilon < 1$ . Give a randomized algorithm that produces a low-dimensional mapping  $f : V \rightarrow \mathbb{R}^T$  such that the areas of all triangles formed from the  $n$  points are preserved with multiplicative error  $\epsilon$ . What is the value of  $T$  for your mapping? Please give the exact number and do not use big O notation.

(Hint: If two triangles lie in the same plane (a 2-dimensional affine space) in  $\mathbb{R}^n$ , then under a linear mapping their areas have the same multiplicative error. For every triangle, add an extra point to form a right-angled isosceles triangle in the same plane.)

3. **(20 points) VC-dimension of Axis-aligned rectangles.**
- (a) Prove that no 5 points on the plane  $\mathbb{R}^2$  can be shattered by the class  $C$  of axis-aligned rectangles (that map points inside a rectangle 1 and otherwise 0).
  - (b) Compute the VC-dimension of the class  $C_k$  of  $k$ -dimensional axis-aligned rectangles in  $\mathbb{R}^k$ . In particular, you need to find a number  $d$  such that there exist  $d$  points in  $\mathbb{R}^k$  that can be shattered by the  $C_k$ , and prove that any  $d + 1$  points in  $\mathbb{R}^k$  cannot be shattered by  $C_k$ .
4. **(10 points) Conditional Expectation.** Suppose  $Y : \Omega \rightarrow \mathbb{R}$  is a random variable and  $W : \Omega \rightarrow \mathcal{U}$  is a random object defined on the same probability space  $(\Omega, \mathcal{F}, Pr)$ . Prove that  $E[Y] = E[E[Y|W]]$ . You may assume that both  $\Omega$  and  $\mathcal{U}$  are finite.
5. **(20 points) Using  $\epsilon$ -Nets for Learning.** Suppose  $X$  is a set with some underlying distribution  $D$  and  $C$  is a class of boolean functions on  $X$ , and the VC-dimension of  $(X, C)$  is  $d$ . Moreover, suppose there is some function  $F_0 \in C$  that corresponds to some classifier that we wish to learn. The model we have is that we can sample a random  $x \in X$  and ask for the value  $F_0(x)$ . After seeing  $m$  such samples  $S$  in  $X$ , we pick a function  $F_1 \in C$  that agrees with  $F_0$  on  $S$ . The hope is that  $F_1$  and  $F_0$  would agree on most points in  $X$  (according to distribution  $D$ ).
- (a) Define another class  $C'$  of boolean functions on  $X$  such that if  $S$  is an  $\epsilon$ -net under  $C'$ , and  $F \in C$  is a function that disagrees with  $F_0$  on more than  $\epsilon$  fraction (weighted according to  $D$ ) of points in  $X$ , then there exists some  $x \in S$  such that  $F(x) \neq F_0(x)$ . Prove the VC-dimension of  $(X, C')$  for the class  $C'$  that you have constructed.
  - (b) How many samples are enough such that with probability at least  $1 - \delta$  the function  $F_1$  returned disagrees with  $F_0$  on at most  $\epsilon$  weighted fraction of points in  $X$ ?
6. **(75 points)  $\epsilon$ -Sample for  $(X, C)$  with VC-dimension  $d$ .** Suppose  $X$  is a set and  $C$  is a collection of boolean functions such that  $(X, C)$  has VC-dimension  $d$ . In this question, we derive a sufficient number  $m$  of independent random samples from  $X$  with distribution  $D$  such that the resulting bag  $S$  is an  $\epsilon$ -sample under class  $C$  of boolean functions with probability at least  $1 - \delta$ .
- (a) **Introducing Extra Randomness.** (15 points) Suppose we sample  $2m$  copies independently from  $X$  to form the bag  $W$ . Then, we pick  $m$  copies out of  $W$  at random to form  $S$ . In other words,  $W$  can be view as a point in  $X^{2m}$ , and we pick  $m$  distinct coordinates at random and use them to form  $S$ .  
 Let  $A$  be the event that there exists some  $F \in C$  such that  $|E_X[F(x)] - E_S[F(x)]| > \epsilon$ .  
 Let  $B$  be the event that there exists some  $F \in C$  such that  $|E_X[F(x)] - E_S[F(x)]| > \epsilon$  and  $|E_W[F(x)] - E_S[F(x)]| > \frac{\epsilon}{4}$ .

Prove that  $Pr[A] \leq 2Pr[B]$ .

(Hint: Show that  $Pr[\overline{B}|A] \leq \frac{1}{2}$ .)

Observe that given  $A$ , the event  $\overline{B}$  implies that there is some  $F_0 \in C$  such that  $|E_X[F_0(x)] - E_S[F_0(x)]| > \epsilon$  and  $|E_W[F_0(x)] - E_S[F_0(x)]| \leq \frac{\epsilon}{4}$ . This means that  $|E_X[F_0(x)] - E_{W \setminus S}[F_0(x)]| > \frac{\epsilon}{2}$ .

Use Hoeffding's Inequality and you may assume  $m \geq \frac{2 \ln 4}{\epsilon^2}$ .)

- (b) **Conditional Probability.** (10 points) For  $F \in C$ , define  $B_F$  to be the event that  $|E_X[F(x)] - E_S[F(x)]| > \epsilon$  and  $|E_W[F(x)] - E_S[F(x)]| > \frac{\epsilon}{4}$ . (Hence,  $B = \cup_F B_F$ .)

Fix  $F \in C$ . Define  $H_F$  to be the event that  $|E_W[F(x)] - E_S[F(x)]| > \frac{\epsilon}{4}$ . Then, clearly  $B_F \subseteq H_F$ , and so  $Pr[B_F|W] \leq Pr[H_F|W]$ . We analyze  $Pr[H_F|W]$ .

Suppose  $P_{max} := \max_{F \in C} Pr[H_F|W]$ . Prove that  $Pr[B] \leq (2m)^d \cdot P_{max}$ .

(Hint: Recall that  $(X, C)$  has VC-dimension  $d$ . After conditioning on  $W$  which has only  $2m$  points, how many boolean functions can the class  $C$  induce on  $W$ ?)

- (c) **Bounding  $P_{max}$ .** (40 points) This is the most technical part of the proof and this part differs the most from the proof for  $\epsilon$ -net.

After  $W$  and  $F$  are fixed, we know precisely how many copies in  $W$  are marked 1 by  $F$ . Let this number be  $L$ . The only randomness left is the choice of  $S$  among  $W$ . Recall that  $S$  is formed from  $W$  by choosing  $m$  copies from the  $2m$  copies in  $W$ .

We can order the objects in  $W$  in an arbitrary list, and assign one by one whether each object is in  $S$  in the following way: suppose when object  $a$  is considered, there are already  $x$  objects assigned to  $S$  and  $y$  objects assigned to  $W \setminus S$ . Then, object  $a$  is assigned to  $S$  with probability  $\frac{m-x}{(m-x)+(m-y)}$  and to  $W \setminus S$  with probability  $\frac{m-y}{(m-x)+(m-y)}$ .

- i. Suppose the  $L$  objects marked 1 are being considered first. For  $1 \leq i \leq L$ , let  $u_i$  be the variable that takes value 1 if the  $i$ th object is assigned to  $S$  and  $-1$  if it is assigned to  $W \setminus S$ .

Define  $U_i := \sum_{j=1}^i u_j$ . Compute the probability that the  $(i+1)$ st object is assigned to  $S$  in terms of  $i$  and  $U_i$ .

What does it mean when  $U_i > 0$ ? When  $U_i > 0$ , what happens to this probability?

Are the  $u_i$ 's independent?

- ii. Find an expression  $\beta$  in terms of  $\epsilon$  and  $m$  such that  $|E_W[F(x)] - E_S[F(x)]| > \frac{\epsilon}{4}$  iff  $U_L^2 > \beta$ .

(We want to obtain an upper bound for  $Pr[U_L^2 > \beta]$ .)

- iii. We saw that the  $u_i$ 's are not independent. This makes the analysis difficult. Hence, we would like to compare the  $u_i$ 's with another collection of independent random variables. For each  $1 \leq i \leq L$ , we define independent random variable  $\gamma_i$  that takes values in  $\{-1, 1\}$  uniformly, i.e., each value with probability  $\frac{1}{2}$ . Define  $Y_i := \sum_{1 \leq j \leq i} \gamma_j$ .

Observe that we would like  $U_L^2$  to be small. Can you explain intuitively why  $Y_L^2$  is more likely to be larger than  $U_L^2$ ?

Prove that  $E[U_L^2] \leq E[Y_L^2]$ .

(Hint: Prove by induction on  $i$  that  $E[U_i^2] \leq E[Y_i^2]$ . In the inductive step, you might find considering the conditional expectation  $E[U_i u_{i+1} | U_i]$  useful.)

(Optional: Prove that for all non-negative integers  $r$ ,  $E[U_L^{2r}] \leq E[Y_L^{2r}]$ . You may use this result for later parts of the question.)

iv. Let  $t$  be a positive real number. Prove that  $E[\exp(tU_L^2)] \leq E[\exp(tY_L^2)]$ .

(Hint: Recall the Taylor expansion  $\exp(y) := \sum_{r \geq 0} \frac{y^r}{r!}$ .)

v. By considering moment generating functions, prove an upper bound for  $\Pr[U_L^2 > \beta]$ , and conclude that  $P_{max} \leq 2 \exp(-\frac{\epsilon^2 m}{32})$ .

(Hint: Recall from the lecture on Johnson-Lindenstrauss Lemma, we have  $E[\exp(tY_L^2)] \leq (1 - 2tL)^{-1/2}$ , for  $t < \frac{1}{2L}$ .)

(d) **Wrapping Everything Up.** (10 points) Prove that if  $m \geq \max\{\frac{64}{\epsilon^2} \ln \frac{2}{\delta}, \frac{128d}{\epsilon^2} \ln \frac{32d}{\epsilon^2}\}$ , then with probability at least  $1 - \delta$ , the bag  $S$  is an  $\epsilon$ -sample for  $X$  under class  $C$ .

7. **(15 points) Calculation Involving  $\log^*$ .** In this question, you are asked to complete the details of some calculations.

(a) **Deriving  $K_f$ .** Suppose  $f(t) := 6 \ln(36t^4)$ . Derive a constant  $K_f > 0$  such that  $f(t) \geq t$  implies that  $t \leq K_f$ .

(b) Suppose  $k := \min\{i : f^{(i)}(T) \leq K_f\}$ . Prove that  $k = O(\log^* T)$ .

(Hint: Make use of big-O notation carefully and avoid messy calculations.)

8. **(20 points) Packet Routing in a Graph.** We describe a problem that is closely related to job shop scheduling.

**Problem Instance.** Suppose  $G = (V, E)$  is a directed graph. We are given  $m$  source-sink pairs  $\{(s_j, t_j) : j \in [m]\}$ . We wish to send one data packet from each source to its corresponding sink.

(a) For each  $j \in [m]$ , a packet must be sent from  $s_j$  to  $t_j$  via some specific path  $P_j$ . Each path  $P_j$  is simple: this means each (directed) edge appears at most once in  $P_j$ .

(b) It takes one unit of time for a data packet to be sent through a directed edge. An edge can only allow at most 1 data packet to be sent at any time.

**Goal.** Schedule the packets to be sent in the graph so that the *makespan*, which is the time for the last packet to arrive at its sink, is minimized.

Show that the packet routing problem can be reduced to the job shop scheduling problem. In particular, given an instance of the packet routing problem, construct an instance of the job shop scheduling problem such that there exists a packet schedule with makespan  $T$  iff there exists a job shop schedule with makespan  $T$ .