**Ground Rules:** Do all problems below. Solve them either by yourself or in teams. You do **not** need to hand in any of these.

## Problem 1. Linear Algebra Review.

- (i) A matrix A is diagonalizable if there is a diagonal matrix D and an invertible matrix B such that  $A = B^{-1}DB$ . Prove that if A has n linearly independent eigenvectors, then A is diagonalizable. Your proof should be constructive in that it gives an explicit construction of B and D.
- (ii) If  $A_{n \times n}$  is orthogonally diagonalizable (meaning there exists an orthogonal matrix P such that  $P^{-1}AP$  is diagonal), then

$$A = \sum_{i=1}^{n} \delta_i P_i P_i^{\mathsf{T}},$$

where  $\delta_i$  is the *i*-th diagonal of  $D = P^{-1}AP$  and  $P_i$  is the *i*-th column of P.

**Problem 2.** Weighted Majority Algorithm. Suppose we generalize the "expert learning" scenario as follows. In the t-th iteration, the algorithm produces a probability vector  $\mathbf{p}^{(t)} = \langle p_1^{(t)}, p_2^{(t)}, \dots, p_N^{(t)} \rangle \in \Delta_N$  (instead of committing to an option  $i \in \{1, \dots, N\}$ ). The adversary then reveals a loss vector  $\ell^{(t)} = \langle \ell_1^{(t)}, \ell_2^{(t)}, \dots, \ell_N^{(t)} \rangle \in [-1, 1]^N$ . To this end, the algorithm incurs a loss of  $(\mathbf{p}^{(t)})^{\mathsf{T}} \ell^{(t)}$  for this iteration. Prove that the randomized weighted majority algorithm satisfies the following:

**Theorem:** For a fixed  $\varepsilon \leq 1$ , any sequence of loss vectors  $\langle \ell^{(t)} \rangle_{t=1}^T$ , any time T, and any index  $i \in [N]$ , the randomized weighted majority algorithm—aka. Hedge( $\varepsilon$ )—satisfies

$$\sum_{t=1}^{T} \left( \mathbf{p}^{(t)} \right)^{\top} \ell^{(t)} \leq \sum_{i=1}^{T} \ell_{i}^{(t)} + \varepsilon \cdot T + \frac{\ln N}{\varepsilon}.$$

**Problem 3.** LP Duality. Find the dual of the following linear program:

Maximize: 
$$5x_1 + 7x_2 - 2x_3$$
  
Subj. to  $x_1 + x_2 \le 10$   
 $2x_1 + 5x_3 \le 19$   
 $3x_2 - x_3 \ge 1$   
 $x_1, x_2 \ge 0, x_3 \in \mathbb{R}$ 

- **Problem 4.** Probabilistic Proof. Show that if G is a connected planar graph, then G has at least one vertex with degree at most 5. It is useful to know that Euler's formula for planar graphs implies that  $m \le 3n 6$ . (*Hint:* What is the expected degree of a vertex of G?)
- **Problem 5.** Random Walks. Recall that  $K_n$  is the complete graph of n vertices and  $P_n$  is the path graph on n vertices. More specifically,  $P_n$  is the graph with vertices  $\{1, 2, ..., n\}$  and edges  $1 \leftrightarrow 2, 2 \leftrightarrow 3, ..., (n-1) \leftrightarrow n$ . Similarly, the vertices of  $K_n$  are  $\{1, 2, ..., n\}$ . For each graph, determine the following:
  - (i) the expected time to reach vertex n starting from vertex 1.

(ii) the expected time to reach vertex *n* starting from vertex 1 and coming back to vertex 1.

**Problem 6.** Streaming Algorithms. Median of means is a popular trick in amplifying the sharpness of an estimate. Suppose you wish to estimate a quantity  $\tau$  and you have come up with an algorithm A that returns T such that  $\mathbf{E}[T] = \tau$  and  $\mathbf{E}[(T - \tau)^2] = \beta$ . Using the median of means strategy, one can obtain estimate  $\widehat{T}$ , which is hopefully much sharper than T. How many parallel copies of A do we need so that we can guarantee  $\Pr[|\widehat{T} - \tau| < \varepsilon] \ge 1 - \delta$ ? (Hint: Chebyshev's inequality and Chernoff-Hoeffding)