You can look up material on the web and books, but you cannot look up solutions to the given problems. You can work in groups, but must write up the answers individually, and must write down your collaborators’ names.

**Problem 1: Convex Programming (10pt)**
Which of the following problems are NP-hard?

1. Linear Programming (LP)
2. Integer Linear Programming (ILP) — the constraints are linear constraints, but the variables are constrained to be integers.
3. Mixed Integer Programming (MIP) — only some of the variables are constrained to be integral, whereas the rest are allowed to take on fractional values.
4. Quadratic Programming (QP) — the constraints are quadratic constraints, but the variables are no longer constrained to be integers.

If you claim a problem is NP-hard, you must give a proof. These proofs should either involve reductions from 3SAT, or from problems you have already shown to be NP-hard. (You might have seen the hardness of some of these problems in class, but you cannot use those since the reductions were from Knapsack, TSP, etc.)

**Problem 2: (Polytope Containment (20pt))**
Let \( S = \{ x \in \mathbb{R}^n : Ax \leq b \} \) and \( T = \{ x \in \mathbb{R}^n : Cx \leq d \} \). Assuming that \( S \) and \( T \) are nonempty, describe an algorithm for checking whether \( S \subset T \). (Your algorithm should run in time polynomial in the number of dimensions \( n \), and the number of constraints \( m \).)

**Problem 3: (Approximation Algorithms using Linear Programming (15pt))**
The Set Cover problem consists of a set of elements \( U \) and a family of subsets \( \{ S_1, S_2, \ldots, S_m \} \) with each \( S_i \subseteq U \) and \( \bigcup_i S_i = U \). Each set \( S_i \) has a cost \( C_i \). We consider special instances of Set Cover, where each element is contained in at most \( f \) sets. This problem is still NP-hard, so we don’t know how to solve the integer linear program formulation below in polynomial time. 

\[
\begin{align*}
\min \sum_i C_i x_i & \quad (1) \\
\text{subject to} \quad \sum_{j \in S_i} x_i & \geq 1 \quad \text{for all elements } j \in U \quad (2) \\
& \quad x_i \in \{0, 1\}. \quad (3)
\end{align*}
\]

So instead, you solve the LP relaxation obtained by replacing the integrality constraints (3) by non-negativity constraints.

\[
\begin{align*}
\min \sum_i C_i x_i & \quad (4) \\
\text{subject to} \quad \sum_{j \in S_i} x_i & \geq 1 \quad \text{for all elements } j \in U \quad (5) \\
& \quad x_i \geq 0. \quad (6)
\end{align*}
\]

Let \( Z_{IP} \) be the optimal value of the integer program, \( Z_{LP} \) be the optimal value of the LP relaxation (with the values of the variables in the LP being \( x_i^* \)).

1. Show that \( Z_{LP} \leq Z_{IP} \).
2. Given a positive value \( \tau \), consider the following rounding operation: set \( x_i' = 1 \) if \( x_i^* \geq \tau \), and \( x_i' = 0 \) otherwise. What is the largest value of \( \tau \) (in terms of \( f \)) which will always ensure that \( x' \) is a feasible solution to the integer program.
3. Show that the cost of the solution \( \sum_i C_i x'_i \leq (1/\tau) \times Z_{IP} \), and hence we have a set cover solution whose cost is at most 1/\( \tau \) times the cost of the optimal solution.

**Problem 4: (15pt)**

Consider the following problem

\[
\begin{align*}
\text{maximize} & \quad x_1 + x_2 \\
\text{subject to} & \quad 2x_1 + x_2 \leq 4 \\
& \quad x \geq 0
\end{align*}
\]

Starting with the initial solution \( (x_1, x_2) = (1, 1) \) give the direction of the first step assuming you are using the affine scaling interior-point method.

**Problem 5: Sequence Alignment with Concave costs (Extra Credit)**

Give an \( O(nm \log nm) \) algorithm to find the minimum cost alignment between two strings when the score of a gap of length \( \ell \) is \(-f(\ell)\), where \( f(\cdot) \) is a concave function. (I.e., \( f(x_1)/x_1 \geq f(x_2)/x_2 \) when \( x_1 \leq x_2 \).)