Note: Your TA probably will not cover all the problems. This is totally fine, the discussion worksheets are not designed to be finished in an hour. They are deliberately made long so they can serve as a resource you can use to practice, reinforce, and build upon concepts discussed in lecture, readings, and the homework.

In this class, we care a lot about the runtime of algorithms. However, we don't care too much about concrete performance on small input sizes (most algorithms do well on small inputs). Instead we want to compare the *asymptotic (i.e. long-term)* growth of the runtimes.

Asymptotic Notation: The following are definitions for $\mathcal{O}(\cdot), \Theta(\cdot)$, and $\Omega(\cdot)$:

- $f(n) = \mathcal{O}(g(n))$ if there exists a c > 0 where after large enough $n, f(n) \leq c \cdot g(n)$. (Asymptotically, f grows at most as much as g)
- $f(n) = \Omega(g(n))$ if $g(n) = \mathcal{O}(f(n))$. (Asymptotically, f grows at least as much as g)
- $f(n) = \Theta(g(n))$ if $f(n) = \mathcal{O}(g(n))$ and $g(n) = \mathcal{O}(f(n))$. (Asymptotically, f and g grow at the same rate)

If we compare these definitions to the order on the numbers, \mathcal{O} is a lot like \leq , Ω is a lot like \geq , and Θ is a lot like = (except all are with regard to asymptotic behavior).

1 Asymptotics and Limits

If we would like to prove asymptotic relations instead of just using them, we can use limits.

Asymptotic Limit Rules: If $f(n), g(n) \ge 0$:

• If
$$\lim_{n \to \infty} \frac{f(n)}{g(n)} < \infty$$
, then $f(n) = \mathcal{O}(g(n))$.

- If $\lim_{n\to\infty} \frac{f(n)}{g(n)} = c$, for some c > 0, then $f(n) = \Theta(g(n))$.
- If $\lim_{n\to\infty}\frac{f(n)}{g(n)} > 0$, then $f(n) = \Omega(g(n))$.

Note that these are all sufficient (and not necessary) conditions involving limits, and are not true definitions of \mathcal{O} , Θ , and Ω . We highly recommend checking on your own that these statements are correct!)

(a) Prove that $n^3 = \mathcal{O}(n^4)$.

(b) Find an
$$f(n), g(n) \ge 0$$
 such that $f(n) = \mathcal{O}(g(n))$, yet $\lim_{n \to \infty} \frac{f(n)}{g(n)} \ne 0$.

(c) Prove that for any c > 0, we have $\log n = \mathcal{O}(n^c)$. *Hint:* Use L'Hôpital's rule: If $\lim_{n \to \infty} f(n) = \lim_{n \to \infty} g(n) = \infty$, then $\lim_{n \to \infty} \frac{f(n)}{g(n)} = \lim_{n \to \infty} \frac{f'(n)}{g'(n)}$ (if the RHS exists)

(d) Find an $f(n), g(n) \ge 0$ such that $f(n) = \mathcal{O}(g(n))$, yet $\lim_{n \to \infty} \frac{f(n)}{g(n)}$ does not exist. In this case, you would be unable to use limits to prove $f(n) = \mathcal{O}(g(n))$. Hint: think about oscillating functions!

2 Asymptotic Complexity Comparisons

- (a) Order the following functions so that for all i, j, if f_i comes before f_j in the order then $f_i = O(f_j)$. Do not justify your answers.
 - $f_1(n) = 3^n$
 - $f_2(n) = n^{\frac{1}{3}}$
 - $f_3(n) = 12$
 - $f_4(n) = 2^{\log_2 n}$
 - $f_5(n) = \sqrt{n}$
 - $f_6(n) = 2^n$
 - $f_7(n) = \log_2 n$
 - $f_8(n) = 2^{\sqrt{n}}$
 - $f_9(n) = n^3$

As an answer you may just write the functions as a list, e.g. f_8, f_9, f_1, \ldots

(b) In each of the following, indicate whether f = O(g), $f = \Omega(g)$, or both (in which case $f = \Theta(g)$). **Briefly** justify each of your answers. Recall that in terms of asymptotic growth rate, constant < logarithmic < polynomial < exponential.

	f(n)	g(n)
(i)	$\log_3 n$	$\log_4(n)$
(ii)	$n\log(n^4)$	$n^2 \log(n^3)$
(iii)	\sqrt{n}	$(\log n)^3$
(iv)	$n + \log n$	$n + (\log n)^2$

3

3 Recurrence Relations

Master Theorem: If the recurrence relation is of the form $T(n) = aT\left(\frac{n}{b}\right) + O(n^d)$ for some constants a > 0, b > 1, and $d \ge 0$, then

$$T(n) = \begin{cases} O(n^d) & \text{if } d > \log_b a \\ O(n^d \log n) & \text{if } d = \log_b a \\ O(n^{\log_b a}) & \text{if } d < \log_b a \end{cases}$$

Remember that if the recurrence relation is *not* in the form $T(n) = aT\left(\frac{n}{b}\right) + O(n^d)$, you can't use Master Theorem! If this is the case, you can try the following strategies:

1. "Unraveling" the recurrence relation.

We try to recursively keep on plugging in the smaller subproblems (e.g. T(n/2) or T(n-1)) to T(n) to find a pattern or simply directly compute the entire expression.

2. Draw a tree!

We use a tree representation to count the total number of calls on each subproblem, doing so by summing up the work per level.

3. Change of Variables

We can make a change of variables (kind of like u-substitution for integration) to mold our recurrence into a nicer form to work with.

4. Squeeze

For certain recurrence relations where we can't directly compute the solution, we can indirectly solve it by computing upper and lower bounds for the runtime based on known recurrence relations. In particular, if the upper and lower bounds have the same asymptotic growth, then we have our answer!

5. Squeeze + Guess & Check

If we try using the previous method and can't end up with asymptotically equivalent upper/lower bounds, then we resort to guess-and-checking reasonable runtimes between the bounds to arrive at the solution (look up "The Substitution Method for Solving Recurrences – Brilliant" to see how to do this). For the purposes of this class, if you ever have to resort to using this method, the expression for T(n) will always look "nice."

There are a lot more strategies that are out-of-scope for this class, and if you're curious we highly recommend you to read about them in the following link: http://beta.iiitdm.ac.in/Faculty_Teaching/Sadagopan/pdf/DAA/new/recurrence-relations-V3.pdf.

Solve the following recurrence relations, assuming base cases T(0) = T(1) = 1:

(a)
$$T(n) = 2 \cdot T(n/2) + O(n)$$

(b) T(n) = T(n-1) + n

(c)
$$T(n) = 3 \cdot T(n-2) + 5$$

(d)
$$T(n) = 2 \cdot T(n/2) + O(n \log n)$$

(e)
$$T(n) = 3T(n^{1/3}) + O(\log n)$$

(f) T(n) = T(n-1) + T(n-2)

4 Complex Numbers Review

A complex number is a number that can be written in the rectangular form a + bi (*i* is the imaginary unit, with $i^2 = -1$). The following famous equation (*Euler's formula*) relates the polar form of complex numbers to the rectangular form:

$$re^{i\theta} = r(\cos\theta + i\sin\theta)$$

where $a = r \cos \theta$ and $b = r \sin \theta$. In polar form, $r \ge 0$ represents the distance of the complex number from 0, and θ represents its angle. Note that since $\sin(\theta) = \sin(\theta + 2\pi), \cos(\theta) = \cos(\theta + 2\pi)$, we have $re^{i\theta} = re^{i(\theta + 2\pi)}$ for any r, θ .

The *n*-th roots of unity are the *n* solutions to the equation $\omega^n = 1$. They are given by

$$\omega_k = e^{2\pi i k/n}, \qquad k = 0, 1, 2, \dots, n-1$$

(a) Let x = e^{2πi3/10}, y = e^{2πi5/10} which are two 10-th roots of unity. Compute the product z = x · y. Is this an n-th root of unity for some n? Is it a 10-th root of unity? What happens if x = e^{2πi6/10}, y = e^{2πi7/10}? For all your answers, simplify if possible.

(b) Show that for any *n*-th root of unity $\omega \neq 1$, $\sum_{k=0}^{n-1} \omega^k = 0$, when n > 1. *Hint:* Use the formula for the sum of a geometric series $\sum_{k=0}^{n} \alpha^k = \frac{\alpha^{n+1}-1}{\alpha-1}$. It works for complex numbers too!

(c) (i) Find all ω such that $\omega^2 = -1$.

(ii) Find all ω such that $\omega^4 = -1$.