Warmup

What is the order of growth for the following functions? Answer in terms of Θ (for example, $\Theta(n)$).

```
1.1 def fib_iter(n):
        prev, curr, i = 0, 1, 0
        while i < n:
            prev, curr = curr, prev + curr
            i += 1
        return prev
```

 $\Theta(n)$, since the while loop executes n times with each iteration taking a constant $\Theta(1)$ time.

Video walkthrough

```
def fib_recursive(n):
    if n == 0 or n == 1:
        return n
   else:
        return fib_recursive(n - 1) + fib_recursive(n - 2)
```

 $\Theta(\Phi^n)$, where Φ is the golden ratio. As long as you understand the runtime is exponential in n, we would accept your answer.

Video walkthrough

1.3 Write a function that takes in a a linked list and returns the sum of all its elements. You may assume all elements in 1nk are integers.

```
def sum_nums(lnk):
    .....
    >>> a = Link(1, Link(6, Link(7)))
    >>> sum_nums(a)
    14
    if lnk == Link.empty:
        return 0
    return lnk.first + sum_nums(lnk.rest)
```

2 Orders of Growth

When we talk about the efficiency of a function, we are often interested in the following: as the size of the input grows, how does the runtime of the function change? And what do we mean by "runtime"?

• square(1) requires one primitive operation: * (multiplication). square(100) also requires one. No matter what input n we pass into square, it always takes one operation.

input	function call	return value	number of operations
1	square(1)	$1 \cdot 1$	1
2	square(2)	$2 \cdot 2$	1
:	:	:	i:
100	square(100)	$100 \cdot 100$	1
:	:	:	:
n	square(n)	$n \cdot n$	1

• factorial(1) requires one multiplication, but factorial(100) requires 100 multiplications. As we increase the input size of n, the runtime (number of operations) increases linearly proportional to the input.

input	function call	return value	number of operations
1	factorial(1)	$1 \cdot 1$	1
2	factorial(2)	$2\cdot 1\cdot 1$	2
:	:	:	÷
100	factorial(100)	$100 \cdot 99 \cdots 1 \cdot 1$	100
:	:	i :	i i
n	factorial(n)	$n \cdot (n-1) \cdots 1 \cdot 1$	n

For expressing complexity, we use what is called big Θ (Theta) notation. For example, if we say the running time of a function foo is in $\Theta(n^2)$, we mean that the running time of the process will grow proportionally with the square of the size of the input as it increases to infinity.

- Ignore lower order terms: If a function requires $n^3 + 3n^2 + 5n + 10$ operations with a given input n, then the runtime of this function is $\Theta(n^3)$. As n gets larger, the lower order terms $(10, 5n, \text{ and } 3n^2)$ all become insignificant compared to n^3 .
- **Ignore constants:** If a function requires 5n operations with a given input n, then the runtime of this function is $\Theta(n)$. We are only concerned with how the runtime grows asymptotically with the input, and since 5n is still asymptotically linear; the constant factor does not make a difference in runtime analysis.

Kinds of Growth

Here are some common orders of growth, ranked from no growth to fastest growth:

• $\Theta(1)$ — constant time takes the same amount of time regardless of input size

- $\Theta(\log n)$ logarithmic time
- $\Theta(n)$ linear time
- $\Theta(n \log n)$ linearithmic time
- $\Theta(n^2)$, $\Theta(n^3)$, etc. polynomial time
- $\Theta(2^n)$, $\Theta(3^n)$, etc. exponential time (considered "intractable"; these are really, really horrible)

In addition, some programs will never terminate if they get stuck in an infinite loop.

Questions

What is the order of growth for the following functions?

```
2.1 def sum_of_factorial(n):
    if n == 0:
        return 1
    else:
        return factorial(n) + sum_of_factorial(n - 1)
```

 $\Theta(n^2)$, we will call factorial n times with arguments n, n-1, n-2, ..., 0. The sum from 0 to n is approximately n^2 .

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 $\Theta(\log(n))$, because our while loop iterates at most $\log(n)$ times, due to n being halved in every iteration.

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```
2.3 def mod_7(n):
    if n % 7 == 0:
        return 0
    else:
        return 1 + mod_7(n - 1)
```

 $\Theta(1)$, since at worst it will require 6 recursive calls to reach the base case. So this is $\Theta(6)$, which can be reduced to $\Theta(1)$.

4 Orders of Growth & Linked Lists

```
2.4 def bar(n):
    if n % 2 == 1:
        return n + 1
    return n

def foo(n):
    if n < 1:
        return 2
    if n % 2 == 0:
        return foo(n - 1) + foo(n - 2)
    else:
        return 1 + foo(n - 2)

What is the order of growth of foo(bar(n))?
    Θ(n²)</pre>
```

Linked Lists 3

There are many different implementations of sequences in Python. Today, we'll explore the linked list implementation.

A linked list is either an empty linked list, or a Link object containing a first value and the rest of the linked list.

To check if a linked list is an empty linked list, compare it agains the class attribute Link.empty:

```
if link is Link.empty:
    print('This linked list is empty!')
else:
    print('This linked list is not empty!')
Implementation
class Link:
    empty = ()
    def __init__(self, first, rest=empty):
        assert rest is Link.empty or isinstance(rest, Link)
        self.first = first
        self.rest = rest
    def __repr__(self):
        if self.rest:
            rest_str = ', ' + repr(self.rest)
        else:
            rest_str = ''
        return 'Link({0}{1})'.format(repr(self.first), rest_str)
    @property
    def second(self):
        return self.rest.first
    @second.setter
    def second(self, value):
        self.rest.first = value
    def __str__(self):
        string = '<'
        while self.rest is not Link.empty:
            string += str(self.first) + ' '
            self = self.rest
```

return string + str(self.first) + '>'

def multiply_lnks(lst_of_lnks):

3.1 Write a function that takes in a Python list of linked lists and multiplies them element-wise. It should return a new linked list.

If not all of the Link objects are of equal length, return a linked list whose length is that of the shortest linked list given. You may assume the Link objects are shallow linked lists, and that lst_of_lnks contains at least one linked list.

```
>>> a = Link(2, Link(3, Link(5)))
    >>> b = Link(6, Link(4, Link(2)))
    >>> c = Link(4, Link(1, Link(0, Link(2))))
    >>> p = multiply_lnks([a, b, c])
    >>> p.first
    48
    >>> p.rest.first
    12
    >>> p.rest.rest.rest
    ()
    11 11 11
Recursive solution:
    product = 1
    for lnk in lst_of_lnks:
        if lnk is Link.empty:
            return Link.empty
        product *= lnk.first
    lst_of_lnks_rests = [lnk.rest for lnk in lst_of_lnks]
    return Link(product, multiply_lnks(lst_of_lnks_rests))
```

For our base case, if we detect that any of the lists in the list of Links is empty, we can return the empty linked list as we're not going to multiply anything.

Otherwise, we compute the product of all the firsts in our list of Links. Then, the subproblem we use here is the rest of all the linked lists in our list of Links. Remember that the result of calling multiply_lnks will be a linked list! We'll use the product we've built so far as the first item in the returned Link, and then the result of the recursive call as the rest of that Link.

Iterative solution:

```
head = Link.empty
tail = head
while Link.empty not in lst_of_lnks:
    all_prod = prod([l.first for l in lst_of_lnks])
    if head is Link.empty:
        head = Link(all_prod)
        tail = head
    else:
```

```
tail.rest = Link(all_prod)
    tail = tail.rest
    lst_of_lnks = [l.rest for l in lst_of_lnks]
return head
```

The iterative solution is a bit more involved than the recursive solution. Instead of building the list "backwards" as in the recursive solution (because of the order that the recursive calls result in, the last item in our list will be finished first), we'll build the resulting linked list as we go along.

We use head and tail to track the front and end of the new linked list we're creating. Our stopping condition for the loop is if any of the Links in our list of Links runs out of items.

Finally, there's some special handling for the first item. We need to update both head and tail in that case. Otherwise, we just append to the end of our list using tail, and update tail.

Write a function that takes a sorted linked list of integers and mutates it so that all duplicates are removed.

```
def remove_duplicates(lnk):
    """
    >>> lnk = Link(1, Link(1, Link(1, Link(5)))))
    >>> unique = remove_duplicates(lnk)
    >>> unique
    Link(1, Link(5))
    >> lnk
    Link(1, Link(5))
    """
```

Recursive solution:

```
if lnk is Link.empty or lnk.rest is Link.empty:
    return lnk
elif lnk.first == lnk.rest.first:
    lnk.rest = lnk.rest.rest
    remove_duplicates(lnk)
    return lnk
else:
    remove_duplicates(lnk.rest)
    return lnk
```

For a list of one or no items, there are no duplicates to remove.

Now consider two possible cases:

• If there is a duplicate of the first item, we will find that the first and second items in the list will have the same values (that is, lnk.first == lnk.rest.first). We can confidently state this because we were told that the input linked list is in sorted order, so duplicates are adjacent to each other. We'll remove the second item from the list.

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Finally, it's tempting to recurse on the remainder of the list (lnk.rest), but remember that there could still be more duplicates of the first item in the rest of the list! So we have to recurse on lnk instead. Remember that we have removed an item from the list, so the list is one element smaller than before. Normally, recursing on the same list wouldn't be a valid subproblem.

• Otherwise, there is no duplicate of the first item. We can safely recurse on the remainder of the list.

Iterative solution:

```
cur = lnk
while cur is not Link.empty and cur.rest is not Link.empty:
    if cur.first == cur.rest.first:
        cur.rest = cur.rest.rest
    else:
        cur = cur.rest
return lnk
```

The loop condition guarantees that we have at least one item left to consider with

For each item in the linked list, we pause and remove all adjacent items that have the same value. Once we see that cur.first != cur.rest.first, we can safely advance to the next item. Once again, this takes advantage of the property that our input linked list is sorted.

4 Midterm Review

4.1 Write a function that takes a list and returns a new list that keeps only the evenindexed elements of lst and multiplies them by their corresponding index.

```
def even_weighted(lst):
    """
    >>> x = [1, 2, 3, 4, 5, 6]
    >>> even_weighted(x)
    [0, 6, 20]
    """

    return [_______]

return [i * lst[i] for i in range(len(lst)) if i % 2 == 0]
```

Alternatively, we can take advantage of the step size for range to make sure we only consider even numbered indices:

```
return [i * lst[i] for i in range(0, len(lst), 2)]
```

The key point to note is that instead of iterating over each element in the list, we must instead iterate over the indices of the list. Otherwise, there's no way to tell if we should keep a given element.

One way of solving these problems is to try and write your solution as a for loop first, and then transform it into a list comprehension. The for loop solution might look something like this:

```
result = []
for i in range(len(lst)):
    if i % 2 == 0:
        result.append(i * lst[i])
return result
```

Video walkthrough

The **quicksort** sorting algorithm is an efficient and commonly used algorithm to order the elements of a list. We choose one element of the list to be the **pivot** element and partition the remaining elements into two lists: one of elements less than the pivot and one of elements greater than the pivot. We recursively sort the two lists, which gives us a sorted list of all the elements less than the pivot and all the elements greater than the pivot, which we can then combine with the pivot for a completely sorted list.

First, implement the quicksort_list function. Choose the first element of the list as the pivot. You may assume that all elements are distinct.

```
def quicksort_list(lst):
    """
    >>> quicksort_list([3, 1, 4])
```

```
[1, 3, 4]
  .. .. ..
  pivot = lst[0]
  less = _____
  greater = ______
  return ______
def quicksort_list(lst):
  if len(lst) <= 1:
     return 1st
  pivot = lst[0]
  less = [e for e in lst[1:] if e < pivot]</pre>
  greater = [e for e in lst[1:] if e > pivot]
  return quicksort_list(less) + [pivot] +
        quicksort_list(greater)
```

A list with zero or no elements is already sorted. Otherwise, we follow the procedure outline in the description.

We pick a "pivot" to remove from the list. Then, construct less and greater lists that represent items less than and greater than the pivot. Notice that both of these lists are guaranteed to be smaller than our original list, so we are guaranteed that these are valid subproblems for our recursive call.

The sorted version of the items less than the pivot will all be less than the pivot, so adding pivot to the end of that list maintains a sorted list over all the smaller elements plus the pivot. Finally, the sorted version of the items greater than pivot will be all be greater than the items less than the pivot plus the pivot, so adding that to the end will ensure our list is still sorted. After concatenating all these lists together, we have a sorted version of our original list.

4.3 Write a function that takes in a list and returns the maximum product that can be formed using nonconsecutive elements of the list. The input list will contain only numbers greater than or equal to 1.

```
def max_product(lst):
    """Return the maximum product that can be formed using 1st
    without using any consecutive numbers
    >>> max_product([10,3,1,9,2]) # 10 * 9
    90
    >>> max_product([5,10,5,10,5]) # 5 * 5 * 5
    125
    >>> max_product([])
    1
    11 11 11
    if lst == []:
        return 1
    elif len(lst) == 1: # Base case optional
        return lst[0]
    else:
        return max(max_product(lst[1:]), lst[0]*max_product(lst[2:]))
```

At each step, we choose if we want to include the current number in our product or not:

- If we include the current number, we cannot use the adjacent number.
- If we don't use the current number, we try the adjacent number (and obviously ignore the current number).

The recursive calls represent these two alternate realities. Finally, we pick the one that gives us the largest product.

Video walkthrough

4.4 An **expression tree** is a tree that contains a function for each non-leaf node, which can be either '+' or '*'. All leaves are numbers. Implement eval_tree, which evaluates an expression tree to its value. You may want to use the functions sum and prod, which take a list of numbers and compute the sum and product respectively.

```
def eval_tree(tree):
    """Evaluates an expression tree with functions the root.
    >>> eval_tree(tree(1))
    1
    >>> expr = tree('*', [tree(2), tree(3)])
    >>> eval_tree(expr)
    6
    >>> eval_tree(tree('+', [expr, tree(4), tree(5)]))
    15
    """
```

```
if is_leaf(tree):
    return label(tree)
args = [eval_tree(branch) for branch in branches(tree)]
if label(tree) == '+':
    return sum(args)
else: # label(tree) == '*'
    return prod(args)
```

Leaf values are guaranteed to be a number, so we can just return their label.

Otherwise, we have to evaulate each of the branches and then combine their result using whichever operater we have in our current root.

If you want to try this out yourself, note that prod isn't actually a built-in operator in Python. You can write it yourself using something like the following:

```
def prod(iterable):
   from functools import reduce
   from operator import mul
   return reduce(mul, iterable, 1)
```

Video walkthrough

4.5 Complete redundant_map, which takes a tree t and a function f, and applies f to the node (2^d) times, where d is the depth of the node. The root has a depth of 0.

```
def redundant_map(t, f):
 11 11 11
 >>> double = lambda x: x*2
 >>> tree = Tree(1, [Tree(1), Tree(2, [Tree(1, [Tree(1)])])])
 >>> print_levels(redundant_map(tree, double))
 [2] # 1 * 2 ^ (1); Apply double one time
 [4, 8] # 1 * 2 ^ (2), 2 * 2 ^ (2) ; Apply double two times
 [16] # 1 * 2 \hat{} (2 \hat{} 2); Apply double four times
 [256] # 1 * 2 ^ (2 ^ 3); Apply double eight times
 t.label = _____
 new_f = _____
 t.branches = _____
 return t
def redundant_map(t, f):
 >>> double = lambda x: x*2
 >>> tree = Tree(1, [Tree(1), Tree(2, [Tree(1, [Tree(1)])])])
 >>> print_levels(redundant_map(tree, double))
 [2]
 [4, 8]
 [16]
 [256]
 0.00
 t.label = f(t.label)
 new_f = lambda x: f(f(x))
 t.branches = [redundant_map(branch, new_f) for branch in t.branches]
 return t
```

Every time we recurse, we transform our map function into one that is one level deeper in terms of calls to input function f. To see why this will achieve the result we want, let's look at what happens to some input function f.

- The first call to redundant_map will call f once.
- This means on the second call to redundant_map, we pass in a function g that causes the original f to be called two times.
- On the third call to redundant_map, we pass in a function h that causes g to be called two times. Remember that g calls original f twice, so h will end up calling original f four times.

Therefore, each level will have double the calls to f as the previous level, which

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matches the requirements.