

COMPSCI 201 MIDTERM 1 REFERENCE SHEET

Arrays: stores fixed number of entries of a single type

- `Type[] name = new Type[length]`
- contains primitives or objects (vs objects)
- no pretty print (vs yes for collections)
- `.asList()`, `.toArray()`, List to Array with loop

Strings: hold an Array of characters

- `+=` concatenation $O(n)$ because immutable
- must be copied over, unchanged after creation
- `StringBuilder.append()` = $O(1)$ per char; large capacity until overflow, geometric growth

methods

- dynamic methods: called on a created object, has access to object data and arguments
- static methods: called on the class, only has access to arguments

class: blueprint for objects

- specifies data and operations for a type of object
- `.` operator accesses instance variable or method of this object (on what it's called)
- public (can be accessed by code outside of class) vs private (only inside)

objects: an instance of a class

- must allocate memory and create
- constructor: specifies how to create a new object
- `Object`: default class from which all inherit

equals

- `equals()` `Object` default → checks if two objects are the same object in memory
 - should be overridden to check value
- `==` → checks if two objects are the same object in memory

interfaces

- ADT (Abstract data type): specifies what a ds does (functionality) but not how (implementation)
- interface: specify a set of abstract methods that an implementing class must override and define (ADT)
- implement → `public class NewList implements List`
 - must have at least same methods

collections

- interface, group of objects
- need implementing class at creation
 - List: ordered sequence of values
 - Set: unordered collection of unique values
 - Map: collection that associates keys and values

ArrayList

- ArrayList implements List with Array, by keeping an array with extra space at the end
 - Array out of space → make one twice as large and copies old over
- `get()` = $O(1)$
- `contains()` = $O(n)$, calls `.equals()` on every element
- `size()` = $O(1)$
- `add()`
 - $O(1)$ at end of the list (amortized bc copying)
 - $O(n)$ at beginning of list (has to shift over)
- `remove()` = $O(N)$, caused by having to shift all elements

HashSet

- unordered but very efficient; internally creates a HashMap to store elements of set
- `add()`, `remove()`, `contains()`, `size()` = constant time $O(1)$
- loop over with enhanced for loop (or iterator)

HashMaps

- calculates hash (int) of a key with `.hashCode()` to see where to store values
- puts it in a "bucket" of k,v pairs with that hash index
- different keys can have same hash because we round (%) on value of `.hashCode()`
- `put()` adds to hash index bucket, updates if present = $O(1)$
- `get()` gets value at computed location to get hash bucket iterates over hashBucket keys to find `.equals()` key = $O(1)$
- `containsKey()` checks if location empty = $O(1)$

TreeSet/TreeMap

- ordered set/map, slightly less efficient than HashSet/HashMap
- keeps values (for Set) and keys (for Map) sorted by “natural ordering,” uses special kind of binary tree
- most methods are runtime complexity $O(\log n)$

Big O (asymptotic) analysis

- describes the runtime (# of operations performed) or memory (space for storing variables)
- constant time $O(1)$: doesn’t depend on size of input (arithmetic, comparison, object attributes)
- take biggest term

| Big O | Name | Example |
|----------------|---------------|--|
| $O(2^N)$ | Exponential | Calculate all subsets of a set |
| $O(N^3)$ | Cubic | Multiply $N \times N$ matrices |
| $O(N^2)$ | Quadratic | Loop over all <i>pairs</i> from N things |
| $O(N \log(N))$ | Nearly-linear | Sorting algorithms |
| $O(N)$ | Linear | Loop over N things |
| $O(\log(N))$ | Logarithmic | Binary search a sorted list |
| $O(1)$ | Constant | Addition, array access, etc. |

- complexity of the line, multiply by # of times the line is executed, add complexity over all lines
 - for loop when i increments by 1 to n is generally $O(n)$
 - nested for loops where both i and j increment by 1 to n is $O(n^2)$
 - for loop when i is doubled from 1 to n is $O(\log(n))$
- methods: go from the inside to the outside, e.g. for `outer(inner(n))`
 - calculate some value returned by `inner(n)` and runtime of `inner(n)`
 - calculate the runtime of `outer()` on that value return by `inner(n)`
 - add both runtimes to get runtime complexity of `outer(inner(n))`

SUHA (Simple Uniform Hashing Assumption)

- suppose we hash N pairs to M buckets
- SUHA: any given key $k \in K$ is equally likely to be hashed to one of the M possible values, independently of what values other keys might hash to
- assuming uniform hashing, probability two random keys hash to same bucket is $1/M$
- expected number of pairs per bucket is N/M (very likely true)
- time to get the hash is constant, but time to search over hash index “bucket” by calling `.equals()` on everything will take N/M runtime
- we face memory/runtime tradeoff with N pairs and M buckets, where...
 - if $N \gg M$ – too many pairs in too few buckets, runtime N/M is NOT constant
 - if $M \gg N$ – too many buckets, constant runtime but NOT memory efficient
 - usually “load factor”, or maximum N/M ratio allowed in Java is 0.75, where M is slightly bigger than N , overall constant runtime but reasonable memory usage
 - HashMap/Set resizes if load factor is exceeded