# Week 1: Data structures and data formats

Algorithmic Data Science 2021-22



What is an algorithm?

Why do we need to consider

algorithms carefully when doing data

science?

## Try this....

- 1. set input := your first name
- 2. set *sum* := 0
- 3. for each *letter* in *input* do the following:
  - 1. set code := position of letter in the
     alphabet
  - 2. set *sum* := *sum* + *code*
- 4. output sum

NOTE THAT THIS IS NOT PYTHON CODE! It is pseudo-code — a set of instructions written in a clear way independent of any programming language conventions.

## What is an algorithm?

- Its a method, a recipe or a set of steps for doing something.
- Its a **well-defined procedure** that takes a value or set of values as **input** and produces some value or set of values as **output**.
- An algorithm is more than an interface which specifies functionality or mappings from inputs to output

## Algorithms and interfaces

- One function in an interface may be implemented by a variety of different algorithms
- How else could we have achieved the "letter\_sum" of a name?
- Consider the problem of long division. An interface would specify the required inputs and outputs i.e., the mathematical function, but not how to achieve the output in practice.
- What algorithms for division are there?

e.g. 
$$850 > 16 \times 16$$
  $16, 32, 48, \dots, 848$   $850 > 16 \times 16$   $1, 2, 3, \dots, 53$   $\frac{850}{10} \div 16$  then multiply by 10

## Why do we need to consider algorithms carefully in data science?

- We want to process large volumes of data....
- ... as quickly and cheaply as possible!
- How many seconds, minutes, hours or days will it take to process the data?
- How much of the data do we need in main memory at the same time?
- How much disk storage do we need? How fast can we access it?
- How can we speed processing up? Will adding another 'node' mean processing speeds up or slows down?

## Module key information

11 week core module: 2 hour lecture + 2 hour lab for 10 weeks

Module convenor and main point of contact

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Office Hours: Tues 2-3 (online, Microsoft Teams)
Thurs 4-5 (Chich 1, 110 or online)

Please email me to book office hour slots

- All the info you need is on **Canvas**. Please ask questions, and post interesting module-related material you come across on Canvas.
- For student-to-student chat, Discord: <a href="https://discord.gg/UyWcEFYbq4">https://discord.gg/UyWcEFYbq4</a>

#### Covid-19

https://twitter.com/trishgreenhalgh/status/144 2162256779821060



## Learning objectives

- apply knowledge of standard data structures to the formulation and decomposition of big data
- evaluate choice of computing model and data representation based on estimation and measurement of impact on space and time complexity and communication performance
- understand the fundamental issues and challenges of developing parallel distributed algorithms for big data
- apply appropriate methods to store and retrieve structured big data

## Topics

Week	Who	Topic
1	Barrett	Data structures and data formats
2	Barrett	Algorithmic complexity. Sorting.
3	Barrett	Matrices: Manipulation and computation
4	Wakeman	Processes and concurrency
5	Wakeman	Distributed computation
6	Barrett	Map/reduce
7	Barrett	Similarity analysis
8	Barrett	Clustering, graphs/networks
9	Barrett	Databases, PageRank algorithm
10	Barrett	Databases 2, revision and past exam papers
11		independent study

## Data Types

- Programming languages typically support a number of atomic data types:
  - integer
  - floating point number
  - string
  - character
  - boolean
- a single data item can be stored in a variable:

```
name = "Adam Barrett"
```

 python allows dynamic typing. Types of variables do not have to be declared and can change:

```
name = 25
```

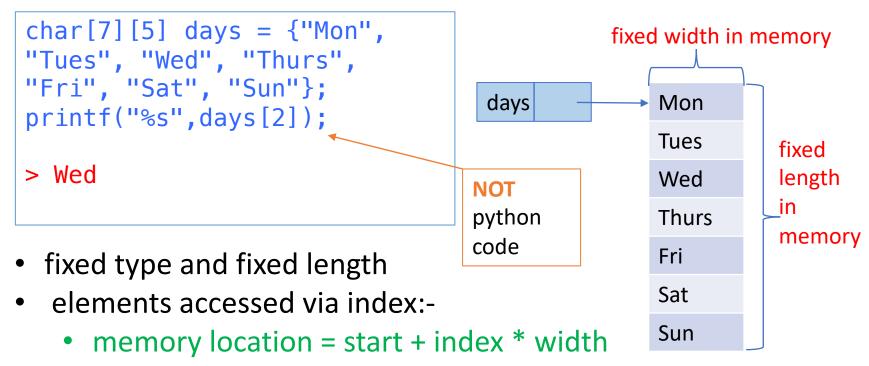
#### Data structures

 A data structure is a collection of data items stored in memory plus a number of operations for manipulating that collection

 Specifies how the data is organised (at least conceptually) and how it should be accessed

## Static arrays

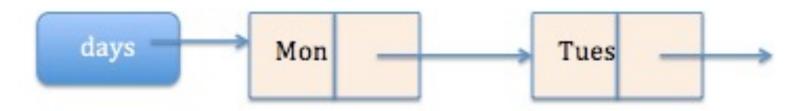
The array is probably the most fundamental of data structures:



- indispensable in C and C++
- In Python: numpy arrays; tuples.

#### Linked lists

• dynamic, grow and change over time



- do not store pointers or indices to every item
- store a link or pointer from each item to the next
- perfect when data is going to be accessed sequentially
- easy to add/append items
- easy to concatenate two lists
- expensive to access i<sup>th</sup> item
  - worst case, follow all n links
  - > O(n)

## Python Lists

code	output	description
planet_list=[]		create a new empty list and store it as planet_list
planet_list=["Mercury", "Venus"]		create a new list with n items and store as planet_list
planet_list.append("Mars")		add an item on to the end of a list
more_planets=["Earth","Jupiter"] bigger_list = planet_list + more_planets		concatenate two lists
print (len(bigger_list))	5	print the length of a list
for p in more_planets: print (p)	Earth Jupiter	iterate over a list and do something with each member
print (bigger_list[2])	Mars	index (or splice) into a list
<pre>print (len(bigger_list[2:]))</pre>	3	splice a list

## Are python lists arrays or linked lists?

- Python lists are (normally) implemented via C routines
- Underlying representation is an array
- Variable length is achieved by over-allocation
- Re-allocate if the array becomes full
- This means fixed cost random access time
  - i.e., time to access *i*<sup>th</sup> element is independent of the length of the list *n*
  - O(1)
- Good list functionality (append, concatenate. . .)

## Maps / Dictionaries

- Not all data is sequential
- Conceptually, a map or dictionary is a collection of (key-> value)
  pairs.
- Might be a set of attributes for an object e.g.,
   adammap = {"name": "adam", "occupation": "lecturer"}
- Or a given attribute for a set of objects e.g.,
   occupations ={"adam": "lecturer", "bob":"porter"}
- Values might be other data structures e.g., lists and dictionaries

## Python dictionaries (dicts)

Code	description
emps={}	new empty dict called emps
<pre>emp={"name": "sue", "occ": "lecturer"}</pre>	new dict with 2 keys called emp
print (emp["name"])	lookup name in emp output: sue
emp["occ"]="reader"	update occupation in emp
emp["age"]=42	create new field in emp
print (emp.keys())	print all of the keys in emp output: ["occ", "age", "name"]
emps[emp["name"]]=emp	store emp in emps with key = name
for (name,record) in emps.items():    if record["age"] > 40:       print (name)	print names of all emps where age > 40 output: sue

## Lookup in a dictionary

- how do you implement a dictionary?
- a list of key-value pairs?

key	value
adam	lecturer
bob	porter
sue	lawyer

- lookup would be really slow
- might have to check every key before you find the right one
- O(n)

## Thought experiment

- We have n folders of information about students (n = 12, n = 10, 000?)
- Each folder is labelled with the student's name
- We have *n* pigeon-holes to store the folders in
- How best do we arrange the folders for fast storage and fast access?

#### Hash tables

- Store the element with key k at slot h(k) where h is a hash function which maps k to a number in the range {0,...,n-1}
- e.g.: h(k) = letter\_sum(k) mod n
- In an ideal world, access time is independent of n, i.e., O(1), and is just dependent on time to compute hash function

#### Collisions

- a hash collision occurs when 2 (or more) keys map to the same slot
- minimise collisions by using a good hash function
- resolve collisions using techniques like chaining or open-addressing

#### Hash functions

- a good hash function satisfies the assumption of simple uniform hashing: each key is equally likely to hash to any of the n slots
- interpret keys as natural numbers
  - e.g., the name "ted" could be interpreted as the triple (116, 101, 100) by looking up the characters in the <u>ASCII</u> <u>character set</u>
  - express sequence using radix-128 notation
     ted = 116 \* 128<sup>2</sup> + 101 \* 128 + 100 = 1913572
- map into *n* slots using e.g., modular division
  - good values for n are primes not too close to exact powers of 2

Optional maths exercise:

Consider a hash function in which  $h(k) = k \mod m$ , where  $m = 2^p - 1$  and k is a character string interpreted in radix  $2^p$ . Show that if string x can be derived from y by permuting its characters, then x and y hash to the same value.

#### Collision resolution

- Chaining resolves collisions by putting all elements that hash to the same slot in a linked list
  - analogous to putting folders that collide in same slot and searching through them at access time
- Open addressing successively probes the hash table until a match or an empty slot is found
  - Hash function takes the probe number as second input.
     E.g.
    - Linear probing:  $h(k,i)=(h_1(k)+i) \mod n$
    - Double hashing:  $h(k, i) = (h_1(k) + i h_2(k)) \mod n$

#### Data formats and data structures

 Here is an extract from the <u>open exoplanet catalogue</u> dataset, which is stored in a csv file

PlanetIdentifier	TypeFlag	PlanetaryMassJpt	RadiusJpt	PeriodDays
KOI-1843.03	0	0.0014	0.054	0.1768913
KOI-1843.01	0		0.114	4.194525
KOI-1843.02	0		0.071	6.356006
Kepler-9 b	0	0.25	0.84	19.22418
Kepler-9 c	0	0.17	0.82	39.03106
Kepler-9 d	0	0.022	0.147	1.592851
GJ 160.2 b	0	0.0321		5.2354
Kepler-566 b	0		0.192	18.42794624

 What Python data structure(s) would you most naturally use when you import a csv file like this? ™ Text ADAMBARRETT935 to 22333 once to join

## What Python data structure would you use to represent a csv file?

List

Dictionary

List of lists

Dictionary of dictionaries

List of dictionaries

Dictionary of lists

Something else

#### csv.reader ≈ list of lists

```
import csv
def readfile(filename):
   with open(filename) as instream:
        csvreader=csv.reader(instream)
        lines=[]
        for line in csyreader:
            lines.append(line)
    return lines
lines=readfile(filename)
print ("{} : {}".format(lines[0][0], lines[0][3]))
print ("{}: {}".format(lines[3][0], lines[3][3]))
```

PlanetIdentifier : RadiusJpt KOI-1843.02 : 0.071

## Representing json

 Here is an extract from items.json file scraped from kaggle in DSRM:

```
[
{"filename": ["perSpindle01.csv", "perSpindle02.csv",
"perSpindle03.csv", "perSpindle4 6.csv", "perTheta.csv"],
"link": "/jbouv27/eeg", "title": "EEG Analysis", "desc":
"Sleep Pattern detection", "popularity": 14, "size": 35387},
{"filename": ["cleanedmrdata.csv"], "link": "/wpncrh/marginal-
revolution-blog-post-data", "title": "Marginal Revolution Blog
Post Data", "desc": "Author Name, Post Title, Word count,
Comment Count, Date, Category Tag", "popularity": 10, "size":
6529129},
```

 What Python data structure(s) would you use to represent a json file like this?

## What Python data structure would you use to represent a json file?

List

Dictionary

List of lists

Dictionary of dictionaries

List of dictionaries

Dictionary of lists

Something else

## json ≈ list or dict of dicts

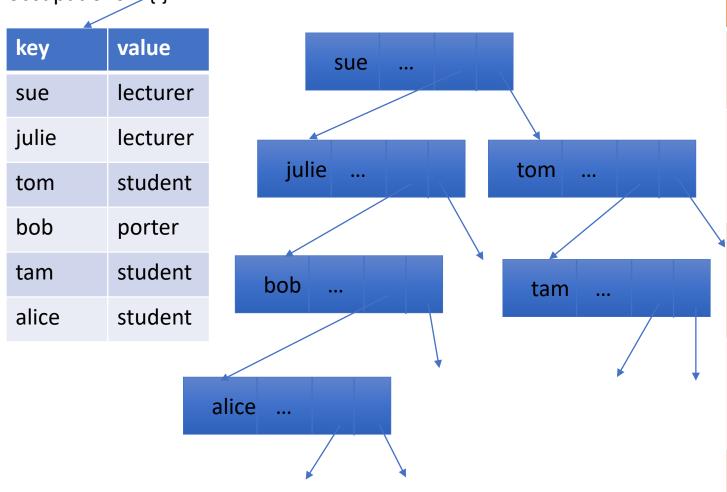
```
import json
def readjsonfile(filename, key='title'):
    with open(filename, 'r') as instream:
        data=json.load(instream)
    datadict={}
    for item in data:
        #what if key does not exist
        #for item or is repeated?
        datadict[item[key]]=item
    return datadict
jsonfile="items.json"
key='title'
data=readjsonfile(jsonfile,key)
a='Pokemon with stats'
b='popularity'
print("{} of {} is {}".format
      (b,a,data['Pokemon with stats']['popularity']))
```

popularity of Pokemon with stats is 2310

# Which format is more compact, csv or json?



occupations = {}

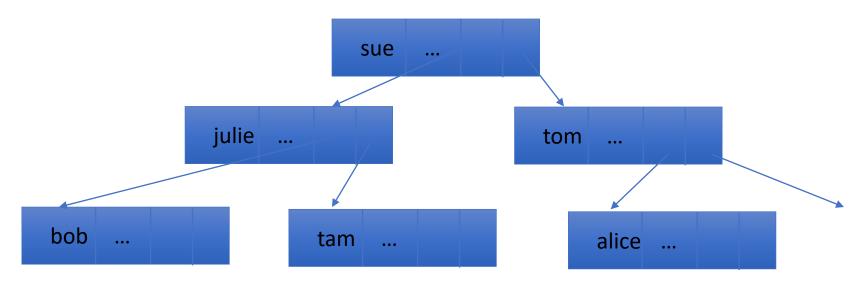


	height	nodes	capacity
	0	1	1
	1	2	3
*	2	4	7
	3	8	15
	h	2 <sup>h</sup>	2 <sup>h+1</sup> -1

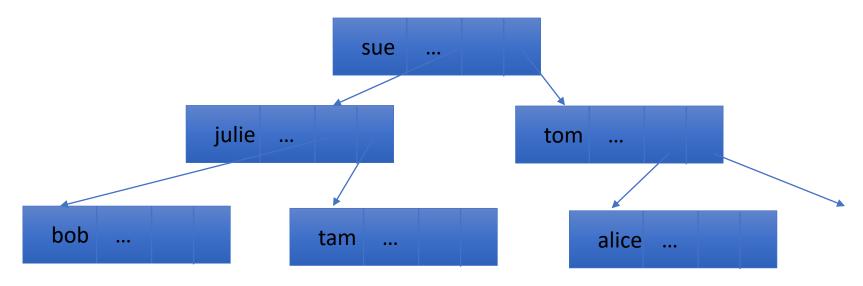
## Binary Search Trees

- balanced tree -> number of nodes at height h is 2<sup>h</sup>
- capacity  $n = \sum_{i=0}^{h} 2^i = 2^{h+1} 1$
- query runs in  $O(\log_2 n)$  time
- unbalanced tree will result if the keys are presorted -> linked list and query runs in O(n) time
- useful data structure when there is a strict order over the keys but keys will be presented in random order
- can be used to implement dictionaries

 Heaps are complete binary trees which satisfy the heap property: Value[parent(i)] ≥ Value[i]

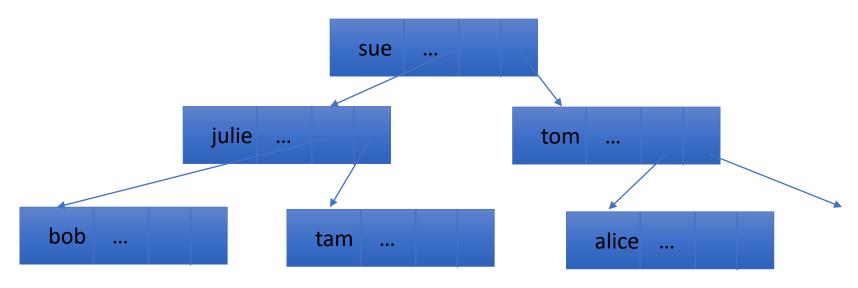


• This is a complete binary tree. Does it satisfy the heap property?

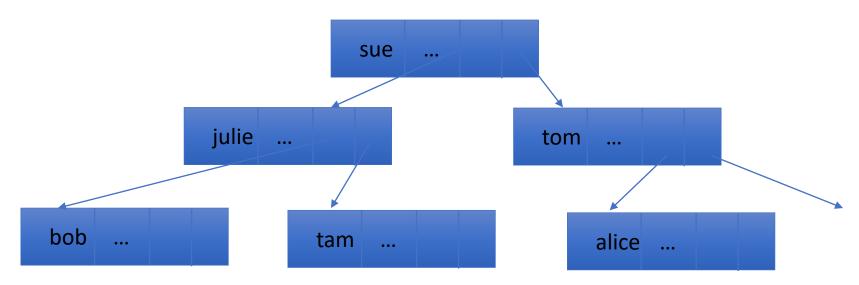


- Run heapify on all non-leaf nodes in a bottom-up fashion i.e.
  - heapify([tom, julie, sue])

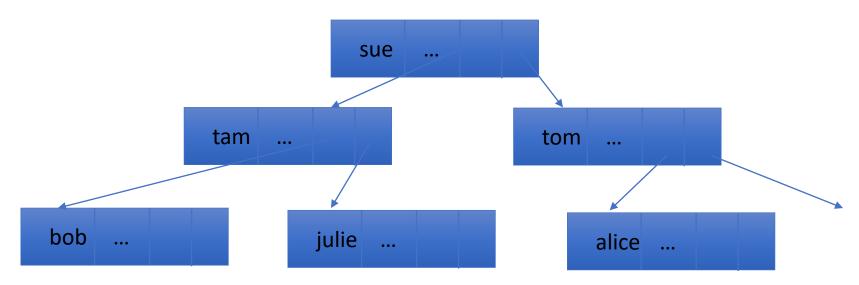
 Heaps are complete binary trees which satisfy the heap property: Value[parent(i)] ≥ Value[i]



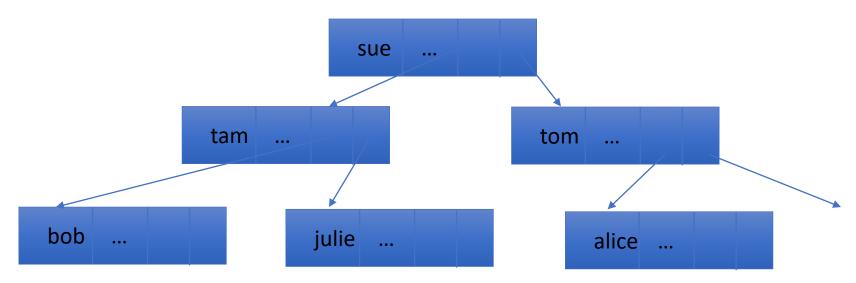
heapify(tom):- tom > alice so heap property satisfied.



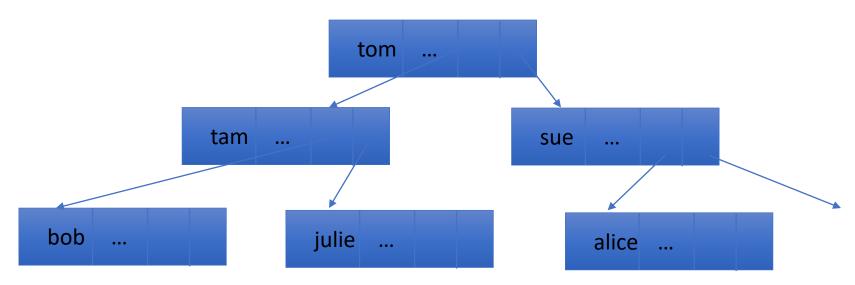
- heapify(julie)
  - 1. tam > julie so swap these nodes
  - 2. heapify(julie)



- heapify(julie)
  - 1. tam > julie so swap these nodes
  - 2. heapify(julie)

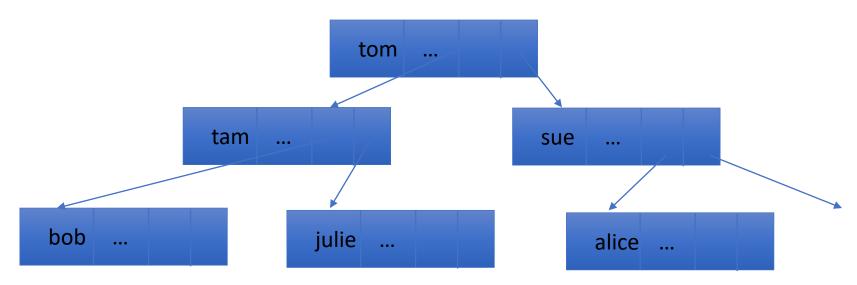


- heapify(sue):-
  - 1. tom > sue & tam so swap sue and tom
  - 2. heapify(sue)



- heapify(sue):-
  - 1. tom > sue & tam so swap sue and tom
  - 2. heapify(sue)

 Heaps are complete binary trees which satisfy the heap property: Value[parent(i)] ≥ Value[i]



heapify(sue): - sue > alice so heap property satisfied

## Why are heaps useful?

- Can be used for sorting (the Heapsort algorithm)
  - Remove maximum element at root of heap
  - Take last element, place it at the root and then heapify
  - Repeat
- Efficient implementation of a 'priority queue'

O(n log n) to build a heap

## Summary

 We have talked about what an algorithm is and why data scientists should care about algorithms. We have introduced a number of important data structures including arrays, linked lists, python lists, dictionaries, hash tables, binary search trees and heaps.