Algorithms for Data Science (Part I - Data

Structures)

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Lecture 08.2.1 (v1.0.1)

Signposting

- This lecture 8.2 of Algorithms for Data Science follows 8.1 on Analysing Algorithms
 - It is about some key algorithms that make Data Science approachable, even without a Big Data Platform.
 - These ideas are building blocks for statistical and machine-learning approaches for inference.
- The lecture is in two parts:
 - ► Part I Data Structures
 - Part 2 Algorithms
- ► This is Part 1, covering Dynamic Data Structures:
 - Hashing
 - Queues/Stacks
 - Linked Lists
 - Binary Trees/Heaps
 - Hash tables

ILOs

- ILO2 Be able to use and apply basic machine learning tools
- ILO4 Be able to use high throughput computing infrastructure and understand appropriate algorithms
- ILO5 Be able to reason about and conceptually align problems involving real data to appropriate theoretical methods and available methodology to correctly make inferences and decisions

Hash functions

- One of the most important components in good algorithmic design is the hash.
- ightharpoonup Simply, a hash h is a map for h(x)=u with:

$$x \in \mathcal{X} \to u \in \mathcal{U}[0, r).$$

- ightharpoonup i.e., we map each item in the space $\mathcal X$ into the Uniform distribution on the integers $0,\dots,r-1$.
- Each item will always map to the same integer.

Hash examples

- Some simple methods for creating keys from integers.
- ▶ Open DSA Data Structures and Algorithms is a great reference.
- ightharpoonup Modulo r

x % 16 # modulo 16

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Binning (floor function or integer division)

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 Mid-Square method: square the value, use the middle digits in the hash

Hash considerations

- ► There are many choices for a hash function in practice. Considerations include:
- **Randomness.** For many applications (e.g. cryptography) we want no correlation between x and y.
- **Locality**. For other applications (e.g. locality sensitive hashing) we want similar x to produce similar u.
- ▶ **Collisions.** We may wish to reduce collisions on a subset of the potential input space. For example, if $x \in [0, r)$ and $u \in [0, r)$ it is possible to eliminate collisions.
- ► Compute. Hash functions vary in their compute cost.
- ► Families. It is often useful to be able to index a family of hash functions with the same computational cost that return different values.

Data Structures

- Data structures are representations of a set of data
- This representation is particularly important when sets are dynamic, i.e. grow or shrink
- We will perform operations on the set, which will have an associated computation cost
- ► The data structure has an associated space cost
- Making the right choice of data structure is an essential component of data science

Fixed size elementary data structures

- We are familiar with the concepts of:
 - \blacktriangleright Arrays: A segment of memory containing n data of the same type
 - Vectors: Arrays with additional operations defined
 - ▶ Multi-dimensional arrays: Arrays of length $n=n_0\times n_1\times \cdots \times n_k$, with entries specified according to a protocol (e.g. row-wise)
 - Matrices/Tensors: Multidimensional arrays with additional operations defined
- ▶ It is clear that arrays are a fundamental concept!

5 1 5 12	3 1 7	12
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- Stacks: Data are stored in an array using "first in, last out": insertions and deletions occur at the same end
 - ▶ Implemented as a pointer to the last read location
- ▶ Queues: Data are stored in an array using "first in, first out": insertions occur one end, deletions the other
 - Implemented as a pointer to the end (for writing) and start (for reading) that tracks removed items



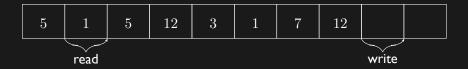
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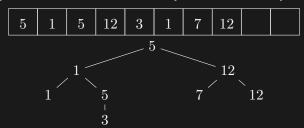
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- Despite implementation similarities, both have different Data Science properties!

Elementary data structures: Linked List



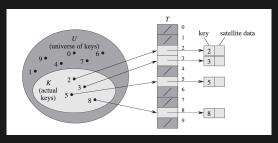
- Linked list: Data are stored in a list, with a pointer to the location of the next item
 - ► Fast traversion, insertion and deletion
 - Slow random access
 - Can be doubly linked

Elementary data structures: Binary Trees & Heaps



- ▶ Binary Trees: Data are stored in a binary linked list, i.e. each node has (up to) two children
 - ▶ Data can be stored at nodes or leaves
 - Critical to define the left/right operation!
- ▶ Position is decided by a key, which can be related to the value ▶ In the picture, values $\leq x$ go left, > x go right
 - Some binary tree structures assign values to internal nodes,
 - e.g. means/ranges
- ► Heaps: A binary tree where each node's key is (larger) than it's children

Elementary data structures: Hash Tables



- Hash Tables: Data location determined by the key
- ▶ The key is a **hash** $x = h_l$: either of an attribute (e.g. a name), or of the value
- ▶ Advantage is O(1) lookup cost. Usage is:
 - I. Compute $u = h_2(x)$
 - 2. Set u' = u%r
 - 3. To insert: store y at this position. On collision, we use some rule to find an empty space, such as rehashing, or storing a linked list.
 - 4. To lookup: retrive this value (using the same rule about collisions).

Signposting

► See 8.2 Part 2 on Algorithms for Data Science

References

- Data structures:
 - ► Cormen et al 2010 Introduction to Algorithms is very accessible and recommended for data structures.
 - ▶ Open DSA Data Structures and Algorithms.