m2 ICA

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0.1 Module 2

0.1.1 ICA: Searches

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```
ICA 1: Linear Search
```

```
[3]: arr = [23, 1, 45, 34, 17]
tgt = 34

def search_linear(arr, tgt):
    n = len(arr)
    for i in range(n):
        if arr[i] == tgt:
            return i
    return -1

search_linear(arr,tgt)
```

[3]: 3

```
pros:
    - simplicity
cons:
```

- no concept of efficiency time complexity: best O(1), worst O(n)

when to use?

- when the array is small in size and the burden of computation is small
- when you don't need to search often or with complexity

ICA 2: Binary Search

```
[15]: arr = [1, 4, 7, 9, 15, 24, 30]
  tgt = 30

def search_binary(arr, tgt):
    # st, end <- arr[0], arr[-1]
    # compute mid between start and end
    # arr[mid] == tgt -> return mid
    # arr[mid] < tgt -> end = mid
```

```
# arr[mid] > tgt -> st = mid

n = len(arr)
st, end = 0, n-1
if arr[st] == tgt: return st
if arr[end] == tgt: return end
while True:
    mid = ((end - st) // 2) + st

if mid == end or mid == st:
    return -1

if arr[mid] == tgt: return mid
elif arr[mid] >= tgt: end = mid
else:
    st = mid

search_binary(arr, tgt)
```

[15]: 6

```
pros:
    - much faster than linear search
cons:
    - requires sorted array
    - more complex to implement than linear search
time complexity: best O(1), worst O(log n)
when to use?
    - when the array is large and sorted
```

- when you need to search often or with complexity

ICA 3: Sort Selection

```
temp = arr[i]
arr[i] = min_val
arr[min_idx] = temp
return arr
sort_selection(arr)
```

[]: [1, 17, 23, 34, 45]

pros:

- easy to implement
- no additional data structures needed
- works well for small datasets

cons:

- not efficient for large datasets
- can be slow for large arrays time complexity: best $O(n^2)$, worst $O(n^2)$
- when to use?
 - when the array is small and unsorted
 - when you need to search for an element in a small dataset

ICA 4: Quick Sort

```
[22]: arr = [12, 4, 5, 6, 7, 3, 1, 15]
      def sort_quick(arr):
          if len(arr) <= 1:</pre>
              return arr
          pivot_idx = len(arr) // 2
          left_arr, right_arr = divide(arr, pivot_idx)
          sorted_left = sort_quick(left_arr)
          sorted_right = sort_quick(right_arr)
          return sorted_left + [arr[pivot_idx]] + sorted_right
      def divide(arr, pivot_idx):
          arr_left = []
          arr_right = []
          n = len(arr)
          for i in range(n):
              if i == pivot_idx:
                  continue
              if arr[i] <= arr[pivot_idx]:</pre>
                  arr_left.append(arr[i])
              else:
                  arr_right.append(arr[i])
```

```
return arr_left, arr_right
sort_quick(arr)
```

[22]: [1, 3, 4, 5, 6, 7, 12, 15]

pros:

- easy to implement
- no additional data structures needed

cons:

time complexity: best $O(n \log n)$, worst $O(n^2)$, avg $O(n \log n)$ when to use?

- if your dataset is large and unsorted
- when you need to sort the array for further operations

ICA 5: Shell Sort description:

- Pick a sequence of gaps (for example, n/2, n/4, ..., 1)
- For each gap, run an insertion pass on the array, comparing and swapping elements that are the defined gap apart
- Cut the gap in half (or to the next sequence value) and repeat until the gap is 1
- Final pass with gap = 1 finishes the sort

pros:

- Large gap passes move elements long distances early so by the time the final pass runs the array is already mostly sorted and requires fewer shifts

cons:

- Performance depends on the chosen gap sequence
- Finding an optimal sequence can be complex and may not yield the best results for all datasets

time complexity:

- Worst case scenario: $O(n^2)$
- Best case scenario: $O(n \log n)$
- Each gap pass n elements O(n), and there are about $O(\log n)$ passes depending on gaps

when to use?

- When working with moderate-sized or partially sorted data and need in-place method faster than n^2
- Useful when memory is limiting factor