

Problems

Sorting

Vergleichsbasierte Sortierverfahren benötigen im schlechtesten Fall und im Mittel mindestens $\Omega(n \log n)$ Schlüsselvergleiche.

Algorithm	Best	Average	Worst	in pl.	stabl.
Bubble	$\Omega(n)$	$\Theta(n^2)$	$O(n^2)$	yes	yes
Selection	$\Omega(n^2)$	$\Theta(n^2)$	$O(n^2)$	yes	no
Insertion	$\Omega(n)$	$\Theta(n^2)$	$O(n^2)$	yes	yes
Merge	$\Omega(n \log n)$	$\Theta(n \log n)$	$O(n \log n)$	poss.	yes
Heap	$\Omega(n \log n)$	$\Theta(n \log n)$	$O(n \log n)$	yes	no
Quick	$\Omega(n \log n)$	$\Theta(n \log n)$	$O(n^2)$	yes	no
Radix	$\Omega(nk)$	$\Theta(nk)$	$O(nk)$		
Bucket	$\Omega(n+k)$	$\Theta(n+k)$	$O(n^2)$		

Bubblesort

1. gehe von links nach rechts durch
2. vergleiche i mit $i + 1$
3. falls $i > i + 1$ tausche
4. wiederhole bis beim durchgang keine vertauschung mehr nötig

Vergleiche: $\Theta(n^2)$

Vertauschungen: worst case: $\Theta(n^2)$

Worst case: "Umgekehrt" sortiert

Selection Sort

1. kleinstes Elementes durch Swap im unsortierten Teil des Arrays
2. tausche kleinstes Element mit erstem in unsort. Teil
3. Unsortierter Teil wird ein Element kleiner wiederhole n mal

Vergleiche: worst case: $\Theta(n^2)$

Vertauschungen: worst case $n - 1 \in \Theta(n)$

Insertion Sort

1. erstes Element rechts vom sort. Teil an richtige stelle im sort. Teil "schieben"
2. so lange bis sort Teil alle Elemente enthält
3. (sort Teil sollte von links nach rechts in jedem schritt wachsen)

Vergleiche: worst case: $\Theta(n \log n)$

Vertauschungen: worst case $(n - 1)^n \in \Theta(n^2)$ (in reverse order)

Shellsort

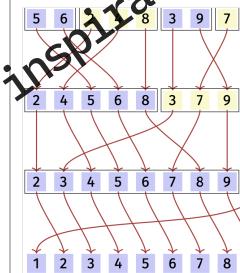
Insertion Sort auf Teilfolgen der Form $(A_{k,i}, i \in \mathbb{N})$ mit absteigenden Abständen k . Letzte Länge ist zwingend $k = 1$. Pratt 1971: $\mathcal{O}(n \log^2 n)$

Mergesort

normal



natural



1. Divide unsorted list into n sublists, each containing one element (trivially sorted)
2. Repeatedly merge sublists to produce new sorted sublists until there is only one sublist remaining (=sorted list)

Vergleiche: worst case: $\Theta(n \log n)$

Vertauschungen: immer (auch natural) $\Theta(n \log n)$

Natürliche 2-Wege Mergesort

same wie Mergesort, aber man nutzt bestehende sortierte Listen "Runs" aus. Asymptotisch nicht besser als regulär mergesort!

Quicksort

1. Pivotieren!
 - (a) von links kommen, stehenbleiben wenn etwas grösser als pivot
 - (b) von rechts kommen, stehenbleiben wenn etwas kleiner als pivot
 - (c) switch diese
 - (d) wenn links und rechts gekreuzt: halt und sort pivot einfügen
 - (e) ein Durchgang abgeschlossen; nächste iteration
2. Quicksort auf Subproblem

worst case: Pivotelemente = extrema

Vergleiche: unwahrscheinlich worst case $\Theta(n^2)$ aber im Mittel $\mathcal{O}(n \log n)$

Vertauschungen: worst case $\mathcal{O}(n \log n)$. Speicherplatzbedarf $O(n)$, aber kann vermieden werden: Rekursion nur auf dem kleineren Teil. Dann garantiert $\mathcal{O}(\log n)$ Rekursionsstiefe und Speicherplatzbedarf.

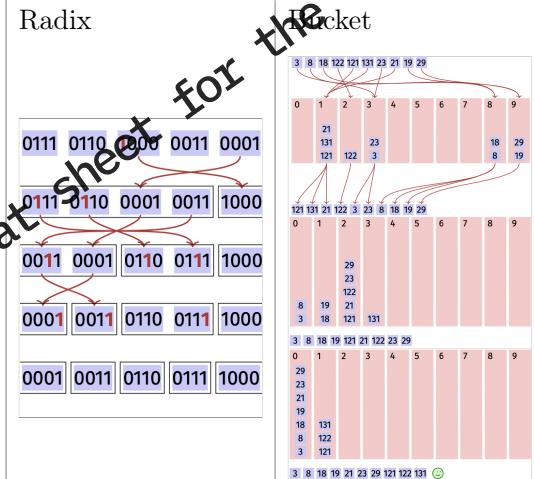
Heapsort

Bucket Sort

1. Verteilung der Elemente auf die Buckets nach Ziffer i
2. Jeder Bucket wird mit einem weiteren Sortierverfahren (Mergesort) sortiert
3. Der Inhalt der sortierten Buckets wird konkateneriert
4. Zifferanzahl-mal wiederholen

Radix vs. Bucket

Radix



The initial pass of both is exactly the same. The elements are put in buckets of incremental ranges depending on the number of digits in the largest number. In the next pass, BucketSort orders up these buckets and appends them into one array. RadixSort appends the buckets without further sorting and re-buckets it based on the second digit of the numbers. BucketSort is more efficient for Dense arrays, while RadixSort can handle sparse arrays well.

Selection (+Median)

The k 'th element of a sequence of n elements can (worst-case) be found in $\Theta(n)$ steps. Special case $k = n/2$: median.

Ranking

- | | |
|-------------------------|-----------------------------|
| $\mathcal{O}(n^2)$ | Repeatedly find minimum |
| $\mathcal{O}(n \log n)$ | Sort, then choose $A[k]$ |
| $\mathcal{O}(n)$ exp. | Quickselect with random piv |
| $\mathcal{O}(n)$ worst | Median of Medians (Blum) |

Search

- The binary sorted search algorithm requires $\Theta(\log n)$ fundamental operations
- Any comparison-based *search* algo on *sorted* data with length n requires in the worst case $\Omega(\log n)$ comparisons
- Any comparison-based *search* algo with *unsorted* data of length n requires in the worst case $\Omega(n)$ comparisons

Abstract Data Types and Implementations**List**

Usage

- Insertion (begin/middle/end to be considered sperately when implementing!)
- traversal in one direction

Implementation

- Linked Lists (insert at begin/middle)
- Arrays (insert at end)

Queue

Usage

- insert (at end)
- traverse in order of insertion
- FIFO

Implementation

- (doubly) linked list

	enqueue	delete	search	concat
(A)	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$	$\Theta(n)$
(B)	$\Theta(1)$	$\Theta(n)$	$\Theta(n)$	$\Theta(1)$
(C)	$\Theta(1)$	$\Theta(1)$	$\Theta(n)$	$\Theta(1)$
(D)	$\Theta(1)$	$\Theta(1)$	$\Theta(n)$	$\Theta(1)$

(A) = Einfach verkettet
(B) = Einfach verkettet, mit Dummyelement am Anfang und Ende
(C) = Einfach verkettet, mit einfache indirekter Elementadressierung
(D) = Doppelt verkettet

Priority Queue ("Heap")

Usage

- insert elems
- extraction of some highest/lowest elem (for some specified order ("priority"))
- NOT actually SORTED

Implementation

- min/max-Heap

Stack

Usage

- insert at begin
- remove from begin
- LIFO

Implementation

- singly linked list

Dictionary

Usage

- elem insertion and removal
- search of elems

Implementation

- Hash Tables (unordered mapping)
- Balanced Trees (AVL) (ordered mapping)

ADT zur Verwaltung von Schlüsseln aus \mathcal{K} mit Operationen

- insert(k, D):** Hinzufügen von $k \in \mathcal{K}$ in Wörterbuch D . Bereits vorhanden \Rightarrow Fehlermeldung.
- delete(k, D):** Löschen von k aus dem Wörterbuch D . Nicht vorhanden \Rightarrow Fehlermeldung.
- search(k, D):** Liefert true wenn $k \in D$, sonst false.

Set

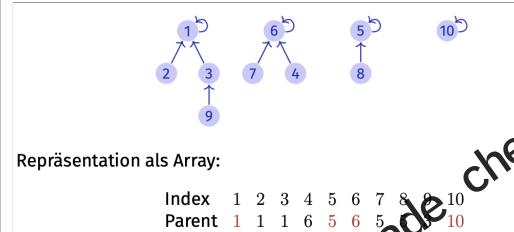
Usage

- Representing math sets
- insert
- check if in same set

Implementation

- Union-find datastruct

- find() worst case runtime $\mathcal{O}(n)$ optimized ($\mathcal{O}(\log(n))$)
- all other virtually $\mathcal{O}(\alpha(n))$, α : inv. Ackermann-f()

**Hash Tables****Hashing****Vocabulary****Prehashing**

$\mathcal{K} \rightarrow \mathbb{N}$. i.e. mapping keys onto integers for further use

Collision

$h(k_i) = h(k_j) i \neq j$. i.e. hash function maps two different keys onto same integer

- Chaining:** Store all $h(k_i) = h(k_j) i \neq j$ in one (worst case very long) linked list. Positive: can overcommit (more entries than slots) and easy to remove entries. Negative: Memory consumption of the chains. Alternative: Closed hashing with open addressing

Simple Uniform Hashing

each key is equally likely to hash to any of the m slots, independently of where any other key has hashed to

Uniform Hashing

the probing sequence of each key is equally likely to be any of the $m!$ permutations of the possible sequences over the hash table of size m

Open Addressing

Position in hash table is not fixed and depends on previous entries

Closed Hashing

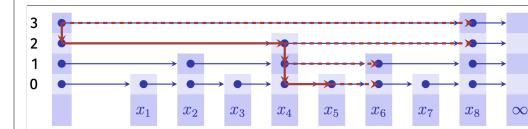
Entries stays in table

Probing

m : Table size

k : Prehashed value

j : j 'th step in probing sequence starting at $j = 0$ and from the very first hash on!



• Linear Probing

$$s(k, j) = h(k) + j$$

Im Fall einer Kollision einfach linear in die vorgegebene Richtung "wandern" bis man auf einen freien Slot kommt

• Quadratic Probing

$$s(k, j) = h(k) + \left\lceil \frac{j}{2} \right\rceil^2 \cdot (-1)^{j+1}$$

Bei Koll.: jeweils j^2 für jeden Wert j^2 erst in pos. dann in neg. Richtung springen und so weiter bis freier Slot auftaucht

• Double Hashing

$$s(k, j) = h(k) + j \cdot h'(k)$$

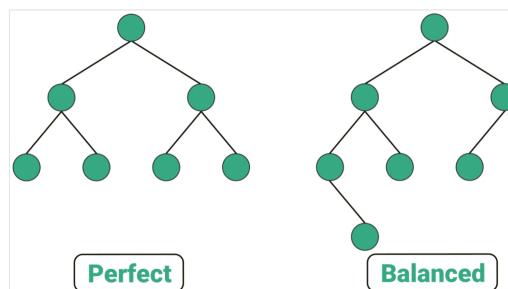
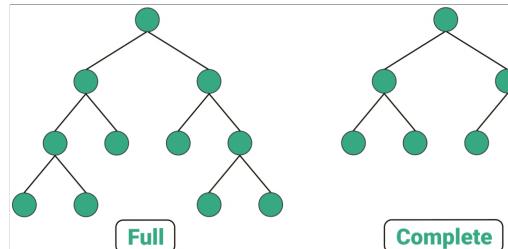
Basically linear probing, nur hält nicht in 1-Schritten sondern in $h'(k)$ -Schritten. **Tipp:** Ganz am Anfang $s(k, j)$ explizit ausschreiben!

Trees

Traversing $\begin{cases} \text{pre-order: } n, n.L, n.R \\ \text{in-order: } n.L, n, n.R \\ \text{post-order: } n.L, n.R, n \end{cases}$

Overview

Suchbäume	Heaps	Balancierte Bäume
	Min-Heap / Max-Heap, std::map	AVL, red-black tree
in C++:	std::map::map	std::map
Einfügen	$\Theta(h(T))$	$\Theta(\log n)$
Suchen	$\Theta(n)$ (!)	$\Theta(\log n)$
Löschen	$\Theta(h(T))$	Suchen + $\Theta(\log n)$
Bemerkung	$\Theta(\log n) \leq \Theta(h(T)) \leq \Theta(n)$	$\Theta(\log n)$



Binary Search Trees (BST)

Suchbaumeigenschaften

- Jeder Knoten v speichert 1 Schlüssel k
- Schlüssel im linken Teilbaum $v.left$ kleiner als $v.key$
- Schlüssel im rechten Teilbaum $v.right$ grösser als $v.key$

Suchen von k

- falls $v.key = k$: gefunden!
- falls $v.key < k$: $v.key \leftarrow v.l$
- sonst $v.key \leftarrow v.r$
- Repeat

Einfügen von k

- suchen(k)
- falls gef.: bereits drin
- falls nicht: dort inserten

k entfernen ($\mathcal{O}(h(T))$)

- k keine Kinder:**
 k durch Blatt ersetzen
- k ein Kind:**
 k durch Kind ersetzen
- k zwei Kinder:**
 k durch grössten in k.l oder kleinsten in

k.r ersetzen ("symmetrischer Nachfolger")

(Max-)Heaps

- vollständig, bis auf die letzte Ebene
- Lücken des Baumes in der letzten Ebene höchstens rechts
- Heap-Bedingung:** Schlüssel eines Kindes kleiner (grösser) als der des Elternknotens

1-indexed
 $\text{child}(i) = \{2i, 2i+1\}$
 $\text{parent}(i) = \{\lfloor i/2 \rfloor\}$

0-indexed
 $\text{child}(i) = \{2i+1, 2i+2\}$
 $\text{parent}(i) = \{\lfloor (i-1)/2 \rfloor\}$

$H(n) = \lceil \log_2(n+1) \rceil$

$\text{heapify}() \in \mathcal{O}(\log n)$ redo heap prop.

$\text{build_Heap}() \in \mathcal{O}(n)[sic!]$ array into heap

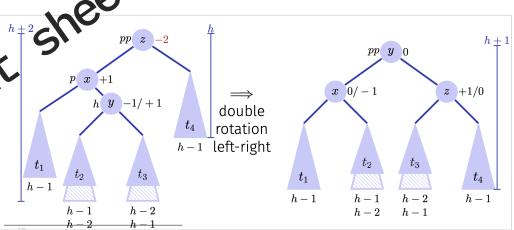
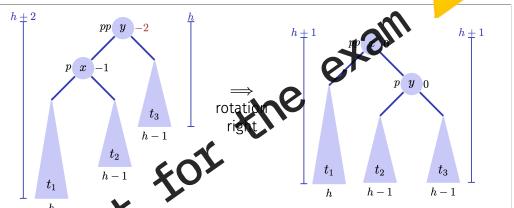
$\text{insert}(v) \in \mathcal{O}(\log n)$ insert at last pos.

redo heap prop. by climbing

$\text{pop_max}() \in \mathcal{O}(\log n)$ switch w/last pos.

redo heap prop. by sinking
/switching w/ ggf. bigr child

(i -th level of a bin. tree has max. 2^i nodes)



(Minimum) Spanning Trees

- kann mit greedy Algos gefunden werden (Kruskal, oder Jarník/Prim/Dijkstra)
- (not unique) tree

Beweise

Induktionsbeweise

- Hypothesis: Definiere $f(n)$
- Base Case: Beweise $f(1) = T(1)$
- Step: $Hyp \implies T(2^{k+1}) = f(2^{k+1})$
Forme $T(2^{k+1})$ mit Hyp um, damit $= f(2^{k+1})$ rauskommt

Beweise mit Invarianten

```
// pre: b>0
// post: return a*b
int f(int a, int b) {
    int res = 0;
    while (b > 0) {
        if (b % 2 != 0) {
            res += a;
            --b;
        }
        a *= 2;
        b /= 2;
    }
    return res;
}
```

let $x := a \cdot b$.
here: $x = a \cdot b + res$
if here $x = a \cdot b + res$...
... then also here $x = a \cdot b + res$
 b even
here: $x = a \cdot b + res$
here: $x = a \cdot b + res$ and $b = 0$
therefore $res = x$.

AVL Trees

AVL-Bedingung: $\forall v$ eines Baumes gilt:

$$bal(v) := h(T_r(v)) - h(T_l(v)) \in \{-1, 0, 1\}$$

Worst-case runtimes of $\mathcal{O}(\log n)$ for searching, insertion and deletion of keys.

Asymptotics **\mathcal{O} -Reihenfolge bestimmen** **\mathcal{O} -Mächtigkeiten**

$$c, \log \log n, \log^c n, \sqrt{n}, n, n \log n, n^c, c^n, n!, n^n$$

$$\binom{n}{k} = \frac{n!}{k!(n-k)!} \in \Theta(n^k) \quad \log(n!) \in \Theta(n \log n) \quad n! \in \mathcal{O}(n^n)$$

Summen

$$\sum_{i=0}^n c^i = \frac{c^{n+1} - 1}{c-1} \in \mathcal{O}(1)$$

$$\sum_{i=0}^n i = \frac{n \cdot (n+1)}{2} \in \mathcal{O}(n^2)$$

$$\sum_{i=0}^n i = \sum_{i=0}^n (n-i) \in \mathcal{O}(n^2)$$

$$\sum_{i=0}^n i^2 = \frac{n(n+1)(2n+2)}{6} \in \mathcal{O}(n^3)$$

$$\sum_{i=0}^n i^k \in \mathcal{O}(n^{k+1})$$

$$\binom{n}{k} = \frac{n!}{k!(n-k)!} \in \mathcal{O}(n^k)$$

$$\sum_{i=0}^n i \in \mathcal{O}(n^4)$$

Logarithmen

$$x = \sqrt[n]{a} \Leftrightarrow x^n = a$$

$$\log_b 1 = 0$$

$$\log_b b = 1$$

$$\log_b(a \cdot c) = \log_b a + \log_b c$$

$$\log_b\left(\frac{a}{c}\right) = \log_b a - \log_b c$$

$$\log_b(a^c) = c \cdot \log_b a$$

$$\log_b a = \frac{\log_c a}{\log_c b}$$

Landaunotation**Standarddefinition**

$$\mathcal{O}(g(n)) = \{ \exists c > 0, \exists n_0 \in \mathbb{N} : \forall n \geq n_0 : 0 \leq f(n) \leq c \cdot g(n) \}$$

$$\Omega(g(n)) = \{ \exists c > 0, \exists n_0 \in \mathbb{N} : \forall n \geq n_0 : 0 \leq c \cdot g(n) \leq f(n) \}$$

$$\Theta(g(n)) = \{ \exists c > 0, \exists n_0 \in \mathbb{N} : \forall n \geq n_0 : 0 \leq \frac{1}{c} \cdot g(n) \leq f(n) \leq c \cdot g(n) \}$$

$$\mathcal{O}(g(n)) \cap \Omega(g(n))$$

Grenzwerte

$$\lim_{n \rightarrow \infty} \left(\frac{f(n)}{g(n)} \right) = \begin{cases} 0 & \Rightarrow f \in \mathcal{O}(g) \\ c & \Rightarrow f \in \Theta(g) \\ \infty & \Rightarrow f \in \Omega(g) \end{cases}$$

Rekursionsgleichungen**Mastertheorem**

$$T(n) = a \cdot T\left(\frac{n}{b}\right) + f(n), \quad (a \geq 1, b > 1)$$

 a : Anzahl Unterprobleme $1/b$: Aufteilungsquotient $f(n)$: Div.- und Summ.kosten

1. Forme Rek.gleichung ins Schema um
2. Berechne $K := \log_b a$
3. Fallunterscheidung machen ($\varepsilon > 0$):

$$f \in \begin{cases} \mathcal{O}(n^{K-\varepsilon}) & \Rightarrow T(n) \in \Theta(n^K) \\ \Theta(n^K) & \Rightarrow T(n) \in \Theta(n^K \log n) \\ \Omega(n^{K+\varepsilon}) \wedge af\left(\frac{n}{b}\right) \leq cf(n), 0 < c < 1 & \Rightarrow T(n) \in \Theta(f(n)) \end{cases}$$

Plug and Chuck

- Expand few times
- Notice patterns (careful with multiplication of $T(n)$)
- Write down explicitly
- Formulate explicit formula $f(n)$
- Prove via induction (starting at $f(1)$)

Aufrufe von $f()$

1. Falls loops: einfach mal rechnen
2. Falls schwer: usually $\Theta(2^n)$
3. Auf Notizpapier einfach stur $g(0), g(1), g(2), g(3), \dots$ rechnen, dabei Rekursion nutzen und Trend erkennen
4. Notfalls einfach Rekglg aufstellen und lösen (siehe Code wie)

Code, damit $f()$ $T(n)$ -mal aufgerufen

$$T(n) = \begin{cases} 2T(n/4) + \log_4(n), & n > 1 \\ 0, & n \leq 1 \end{cases}$$

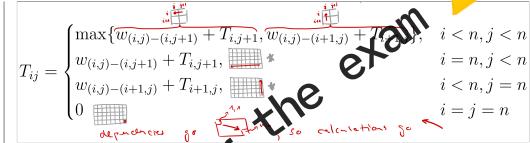
```
if (n > 1){
    g(n/4); g(n/4); // 2T(n/4)
    while(n > 1){ // 0, n <= 1
        f();
        n /= 4; // log_4(n)
    }
}
```

$$T(n) = \begin{cases} 1 + T(n/4), & n > 1 \\ 2, & n = 1 \end{cases}$$

```
if (n == 1){
    f(); f(); // 2 falls n == 1
} else{
    f();
    g(n/4); // T(n/4)
}
```

Dynamic Programming**1. Table with info on subproblems***Dimension of the table?**What are the entries "mathematically"?***2. Computation of base cases***Which entries do not depend on others?***3. Determine computation order***In which order can the entries be computed?***4. Read-out the solution***How can the solution be read out?*

$$\text{Runtime} = \text{Entries} \cdot \frac{\text{Operations}}{\text{Entry}}$$

**Knapsack**

```
int max(int a, int b){
    return (a > b) ? a : b;
}

// Returns the max value that can
// be put in knapsack of capacity W
int knapSack(int W, int wt[], int
val[], int n){
    int i, w;
    vector<vector<int>> K(n + 1,
        vector<int>(W + 1));
    // Build table K[][] in bottom up
    // manner
    for(i = 0; i <= n; i++){
        for(w = 0; w <= W; w++){
            if (i == 0 || w == 0){
                K[i][w] = 0;
            } else if (wt[i - 1] <= w){
                K[i][w] = max(val[i - 1] +
                    K[i - 1][w - wt[i - 1]],
                    K[i - 1][w]);
            } else {
                K[i][w] = K[i - 1][w];
            }
        }
    }
    return K[n][W];
}

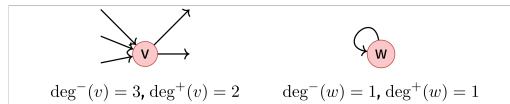
int main(){
    int val[] = { 60, 100, 120 };
    int wt[] = { 10, 20, 30 };
    int W = 50;
    int n = sizeof(val) /
        sizeof(val[0]);
    cout << knapSack(W, wt, val, n);
    return 0;
}
```

This cheat sheet is to be used as inspiration and not as a ready-made cheat sheet for the exam!

Graphs

Notation/Representation

Degrees



For each graph $G(V, E)$ it holds that

- for directed:

$$\sum_{v \in V} \deg^-(v) = \sum_{v \in V} \deg^+(v) = |E|$$

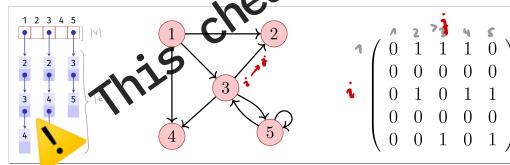
- for undirected Graphs:

$$\sum_{v \in V} \deg(v) = 2|E|$$

Simple Observations

- Generally: $0 \leq |E| \in \mathcal{O}(|V|^2)$
- Connected G : $|E| = \Omega(|V|)$
- Complete G : $|E| = \frac{|V| \cdot (|V|-1)}{2}$
- Maximally:
 - $|E| = |V|^2$ (directed)
 - $|E| = \frac{|V| \cdot (|V|+1)}{2}$ (undirected)
- Also: undirected graphs cannot contain cycles with length 2 (loops have length 1)

Matrix vs. List



Operation	Matrix	List
Find neighbours/successors of $v \in V$	$\Theta(n)$	$\Theta(\deg^+ v)$
find $v \in V$ without neighbour/successor	$\Theta(n^2)$	$\Theta(n)$
$(v, u) \in E$?	$\Theta(1)$	$\Theta(\deg^+ v)$
Insert edge	$\Theta(1)$	$\Theta(1)$
Delete edge (v, u)	$\Theta(1)$	$\Theta(\deg^+ v)$

- Draw ring around all neighbouring vertices from current set
- Draw tree of visited nodes (useful for Qs)
- Move on to next iteration with newly drawn ring/set

Directed Acyclic Graphs (DAG)

a directed graph with *no* directed cycles. A directed graph is a DAG \iff it can be topologically ordered, by arranging the vertices as a linear ordering that is consistent with all edge directions. (\rightarrow TopologicalSort)

Algorithms on Graphs

Depth First Search (DFS)

- (nützlich um Zyklen zu finden)
- **Laufzeit:** $\Theta(|V| + |E|)$
- verfolgt einen Pfad in die Tiefe, bis kein neuer Knoten mehr besucht werden kann
- dann geht zum nächsten Nachbar des Startknotens
- sind alle Pfade, die bei einem Nachbarn beginnen, durchlaufen worden, wird ein anderer Startknoten aus der Menge der unbesuchten Knoten gewählt

Breadth First Search (BFS)

- **Laufzeit:** $\Theta(|V| + |E|)$
- Lösung für kürzester-Weg-Problem wenn Kantengewichte alle uniform

Manual Execution

1. Draw ring around starting vertex s
2. Set $\text{dist}(s) = 0$
3. while(\exists Neighbours) do

Finding MSTs

Kruskal's MST Algo

- $\Theta(|E| \log |V|)$
- nutzt ADT "Union-Find"

Manual Execution

1. Sort edges by weight (min-Heap!)
2. walk through the sorted edges and look at the two nodes that the edge is between
 - if nodes are already in same set (union-find), don't include that edge
 - otherwise we do and now unify those nodes
3. algo terminates when all nodes are unified (i.e. in same set)

Jarnik/Prim/Dijkstra's MST Algo

- Use this one for manual execution!
- $\Theta(|E| \cdot |V|)$ (Fibonacci-heap: $\Theta(|E| + |V| \log |V|)$)
- better than Kruskal on dense graphs (= many edges)
- ähnlich zu Dijkstra's shortest path algo

Manual Execution

1. start with some node as your inner set
2. keep add cheapest edge that is going out from current set until entire graph is in inner set

$$B := A_G^2 = \begin{pmatrix} 0 & 1 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 1 \end{pmatrix}^2 = \begin{pmatrix} 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 2 \\ 0 & 1 & 1 & 1 & 2 \end{pmatrix}$$

Shortest Path Problem

Overview/Rankings

Shortest Path (SP) Algorithms				
	BFS	Dijkstra's	Bellman Ford	Floyd Warshall
Complexity	$O(V+E)$	$O((V+E)\log V)$	$O(VE)$	$O(V^3)$
Recommended graph size	Large	Large/ Medium	Medium/ Small	Small
Good for APSP?	Only works on unweighted graphs	Ok	Bad	Yes
Can detect negative cycles?	No	No	Yes	Yes
SP on graph with weighted edges	Incorrect SP answer	Best algorithm	Works	Bad in general
SP on graph with unweighted edges	Best algorithm	Ok	Bad	Bad in general

Johnson (Solves APSP)

$\mathcal{O}(|V|^2 \log(|V| + |V||E|))$ (with FibHeaps in Dijkstra)

A*

$\mathcal{O}(|E|) = \mathcal{O}(b^d)$

Observations

- negative cycle \Rightarrow no shortest path
- there can be exp. many paths (trying all is futile)
- triangle inequality holds
- subpaths of shortest paths are shortest paths too (optimal substructure)
- shortest paths don't contain cycles

Shortest Path Algorithms

General Algorithm

- Initialise d_s and π_s : $d_s[v] = \infty, \pi_s[v] = \text{null}$ for each $v \in V$
- Set $d_s[s] \leftarrow 0$
- Choose an edge (u, v)
 - Relaxiere (u, v) :
if $d_s[v] > d_s[u] + c(u, v)$ then
 $d_s[v] \leftarrow d_s[u] + c(u, v)$
 $\pi_s[v] \leftarrow u$
- Repeat 3 until nothing can be relaxed any more.
 $\exists u, v \in V \text{ s.t. } d_s[v] \leq d_s[u] + c(u, v) \quad \forall (u, v) \in E$

Dijkstra

- Greedy
- Worst-case $\mathcal{O}(|E| + |V| \log |V|)$ (mit Fibheaps)
- Worst-case $\mathcal{O}((|E|+|V|) \cdot \log |V|)$ (ohne Fibheaps)
- funzt nur mit pos. Kantengewichten
- Grundidee: 3 Mengen (1: kürz. Weg bekannt, 2: direkt anliegend an 1., und 3: noch im Fog of War)

Manual Execution

- Give all vertieces $\text{dist}_{v_i} = \infty$
- Give starting vertex $\text{dist}_s = 0$
- while($t (= \text{endvertex})$ is unmarked) do
 - choose vertex with smallest dist and mark it
 - relax all reachable vertices

Bellman-Ford

- $\mathcal{O}(|V| \cdot |E|)$ with fibheaps
- based on Dynamic Programming (not greedy!)
- used for neg-cycle detection (Arbitrage!)
- funktioniert auch bei negativen Kantengewichten aber nicht bei negativen Zyklen!
- nach $|V|$ Iterationen können noch Relaxierungen gemacht werden \Rightarrow negative Zyklen im Graph (da dass einen shortest path von Länge $|V|$ impliziert, was nicht sein kann)

Manual Execution

```

foreach u ∈ V do
  ds[u] ← ∞; πs[u] ← null // D3 Proj.
ds[s] ← 0;
for i ← 1 to |V| do
  f ← false
  foreach (u, v) ∈ E do
    f ← f ∨ Relax(u, v)
    if f = false then return true // in case of already max. relaxed
  if relaxation was possible in |V|'th iteration,
  return false; // this would imply a negativ cycle
  
```

Floyd-Warschall

- löst all-pairs-shortest path, gibt aber nur path lengths an, nicht paths (paths können aber konstruiert werden)
- funktioniert bei negativen Kanten** aber keine negativen Zyklen!
- best implemented with Adjazenzmatrizen
- can be executed in single matrix (in place)
- also based on Dynamic Programming
- pain in the ass to do by hand since it's $\mathcal{O}(|V|^3)$

Input: Graph $G = (V, E, c)$ without negative weight cycles.

Output: Minimal weights of all paths d

```

d0 ← c
for k ← 1 to |V| do
  for i ← 1 to |V| do
    for j ← 1 to |V| do
      dk(vi, vj) = min{dk-1(vi, vj), dk-1(vi, vk) + dk-1(vk, vj)}
  
```

Johnson

Runs Bellman-Ford when *reweighting* once and then $|V|$ -times Dijkstra on the *reweighted* graph \tilde{G} for shortest path between all pairs of vertices

Reweighting Process, that basically reassigns all path weights c_i a new \tilde{c}_i which is *non-negative*. The shortest path of any graph on which this is done **remains the same**. This is done by:

- add a new vertex s which has paths of cost = 0 to all other vertices
- calculate the shortest paths from s to each of the $v \in V$; call it $d(s, v)$ (with the Bellman-Ford Algo)
- reassign all $c(u, v)$ like this:

$$h(v) := d(s, v)$$

$$\tilde{c}(u, v) := c(u, v) + h(u) - h(v) \geq 0$$

A*

Basically Dijkstra but with a Heuristik (usually Manhattan-distance) (that *underestimates* the actually distance it's going to take)

Input: Positively weighted Graph $G = (V, E, c)$, starting point $s \in V$, end point $t \in V$, estimate $\hat{h}(v) \leq \delta(v, t)$

Output: Existence and value of a shortest path from s to t

```

foreach u ∈ V do
  d[u] ← ∞; f[u] ← ∞; π[u] ← null
d[s] ← 0; f[s] ← h(s); R ← {s}; M ← {}
while R ≠ ∅ do
  u ← ExtractMinf(R); M ← M ∪ {u}
  if u = t then return success
  foreach v ∈ N+(u) with d[v] > d[u] + c(u, v) do
    d[v] ← d[u] + c(u, v); f[v] ← d[v] + h(v); π[v] ← u
    R ← R ∪ {v}; M ← M - {v}
return failure
  
```

Flow Networks G

Flow and Cuts

Flow

A flow $f : V \times V \rightarrow \mathbb{R}$ fulfills:

$$\begin{aligned} \text{Kapazitätsbeschr: } f(u, v) &\leq c(v, u) \\ \text{Symmetrie: } f(u, v) &= -f(v, u) \\ \text{Flusserhaltung: } \sum_{v \in V} f(u, v) &= 0 \\ &\uparrow \forall u \in V \setminus s, t \\ \text{Value: } |f| &= \sum_{v \in V} f(s, v) \end{aligned}$$

- Flow network $G = (V, E, c)$: directed graph with capacities
- Antiparallel edges forbidden: $(u, v) \in E \Rightarrow (v, u) \notin E$.
- Model a missing edge (u, v) by $c(u, v) = 0$.
- Source s and sink t : special nodes. Every node v is on a path between s and t : $s \rightsquigarrow v \rightsquigarrow t$

Cuts

- Cuts are partitions of V into S and T with $s \in S$ and $t \in T$
- Capacity of a cut:

$$c(S, T) = \sum_{v \in S, v' \in T} c(v, v')$$

i.e. Summe der c von den e die gecuttet werden

- minimal cut: cut with minimal capacity
- Flow over a cut:

$$f(S, T) = \sum_{v \in S, v' \in T} f(v, v')$$

i.e. Summe der f von den e die gecuttet werden

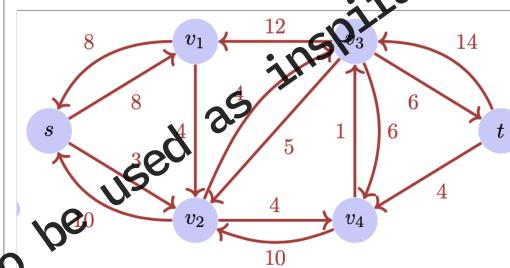
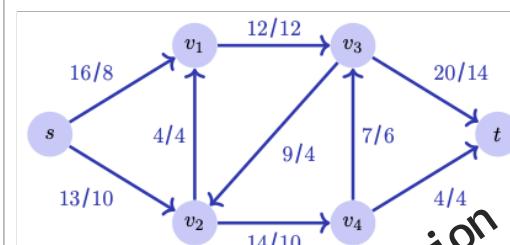
Max-Flow Min-Cut Theorem

Let f be a flow in a flow network $G = (V, E, c)$ with source s and sink t . The following statements are equivalent:

- f is a maximal flow in G
- The rest network G_f does not provide any expansion paths
- It holds that $|f| = c(S, T)$ for a cut (S, T) of G

Rest Network G_f

Definition (c/f)



Rest network G_f provided by the **edges with positive rest capacity**. Weg in Richtung s einfach c von G aber Weg in Richtung t jeweils Restkapazität $c - f$. Rest networks provide the same kind of properties as flow networks with the exception of **permitting antiparallel capacity-edges**.

Theorem 31:

Let $G = (V, E, c)$ be a flow network with source s and sink t and f a flow in G . Let G_f be the corresponding **rest networks** and let f' be a flow in G_f . Then $f \oplus f'$ with

$$(f \oplus f')(u, v) = f(u, v) + f'(u, v)$$

defines a flow in G with value $|f| + |f'|$.

Augmenting Paths

expansion path p : simple path from s to t in the rest network G_f

Rest capacity:

$$c_f(p) = \min\{c_f(u, v), (u, v) = e \in p\}$$

(smallest capacity along path p in G_f)

Flow in G_f

Theorem 32: the map $f_p : V \times V \rightarrow \mathbb{R}$,

$$f_p(u, v) = \begin{cases} c_f(p), & (u, v) \in p \\ -c_f(p), & (v, u) \in p \\ 0, & \text{otherwise} \end{cases}$$

provides a flow in G_f with value $|f_p| = c_f(p) > 0$

Strategy for an Algorithm

With an expansion path p in G_f the flow $f \oplus f_p$ defines a new flow with value

$$|f \oplus f_p| = |f| + |f_p| > |f|$$

Maximal Flow Algorithms

Ford-Fulkerson

- gibt den maximalen Fluss eines Netzes zurück
- anfangs haben alle Kanten den Fluss $f(u, v) = 0$, dann bestimmt man das Restnetzwerk G_f und dadurch einen Erweiterungspfad p der vorgibt, um wieviel man den Fluss erhöhen kann
- might not terminate for $c \in \mathbb{R}$ but always will in $\leq |f_{max}|$ (since always increases by at least 1) steps for $c \in \mathbb{Z}$
- Laufzeit beträgt $\mathcal{O}(f_{max} \cdot |E|)$

Manual Execution

- start with $f(u, v) = 0, \forall v \in V$
- determine rest network G_f and find expansion path p in it
- increase flow along p by the smallest capacity c that lies along p (careful with the direction of flow, sometimes one must decrease it)
- repeat until no more expansion paths p can be found in G_f

```
for (u, v) ∈ E do
    f(u, v) ← 0
while Exists path p : s ↗ t in rest network G_f do
    c_f(p) ← min{c_f(u, v) : (u, v) ∈ p} // smallest c_f along the path p in G_f
    foreach (u, v) ∈ p do
        if (u, v) ∈ E then
            f(u, v) ← f(u, v) + c_f(p)
        else
            f(v, u) ← f(v, u) - c_f(p)
```

Edmonds-Karp

- quasi Ford-Fulkerson, nur wird bei der Wahl des Erweiterungspfades s der kürzeste ausgewählt (z.B. durch BFS)
- $\mathcal{O}(|V| \cdot |E|^2)$ (if $c \in \mathbb{Z}$)
- Number of flow increases $\in \mathcal{O}(|V| \cdot |E|^2)$ (Theorem 34)

Push-Relabel

- has a relaxed flow condition at the beginning and then basically works towards restoring it but in the mean time works with *preflow*
- the residual network G_f therefore works with *preflow* and not with flow itself
- Terminates after $\mathcal{O}(n^2)$ relabels and $\mathcal{O}(n^3)$ pushes
- Runtime $\mathcal{O}(V^2\sqrt{E})$

```

while  $\exists u \in V \setminus \{s, t\} : \alpha_f(u) > 0$  do
  choose  $u$  with  $\alpha_f(u) > 0$  and maximal  $h(u)$ 
  if  $\exists v \in V : c_f(u, v) > 0 \wedge h(v) = h(u) - 1$  then
    push( $u, v$ )
  else
     $h(u) \leftarrow h(u) + 1$ 
  
```

```

if  $\alpha_f(u) > 0$  then
  if  $c_f(u, v) > 0$  in  $G_f$  then
     $\Delta \leftarrow \min\{c_f(u, v), \alpha_f(u)\}$ 
     $f(u, v) \leftarrow f(u, v) + \Delta$ .
  
```

Ford-Fulkerson (conservative)
Invariant: flow conservation
Steps: augmenting paths
Goal: separate s from t in the residual network.

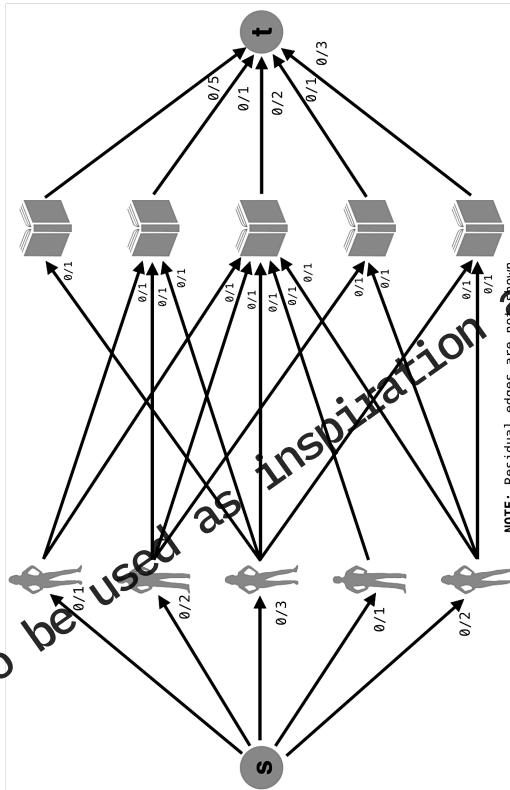
Push-Relabel
Invariant: height invariant (no augmenting path!)
Steps: push flow
Goal: achieve flow conservation

Max Bipartite Matching

- Greedy approach doesn't work. Use maxflow algos (Theorem 35: Ford-Fulkerson) to solve it
- Problems of this kind usually revolve around finding matches between e.g. people and jobs, readers and books, etc.

Example: Readers to books

- Capacity c_s from s to people**
 c books that person can take
- Capacity c_t from books to sink t**
how many times each book can be taken (by all readers combined)
- Edge from a reader to a book**
which books each reader wants (and how many times, even though this doesn't make sense in this example)

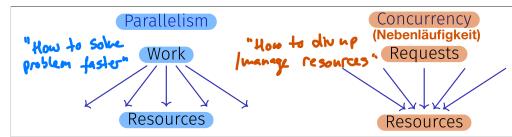
**Node capacity implementation**

Let v be a vertex of G with capacity $c(v)$. Replace G with two vertices v' and v'' and connect them with an edge from v' to v'' with capacity $c(v)$. Replace any incom-

ing edge (u, v) by the edge (u, v') with the same capacity. Further, replace any outgoing edge (v, w) by the edge (v'', w) . By doing this for all vertices of G , we obtain the graph G' and its capacity function c' . A flow in G' can be mapped to a valid flow in G , as the sum over the incoming flow for v' is the same as the sum of the outgoing flow for v'' . Further, this sum never exceeds $c(v)$.

This cheat sheet is to be used as inspiration and not as a ready-made cheat sheet for the exam!

Parallel Programming



Parallelism: use extra resources to solve problem faster

Concurrency: How to div up/manage resources

Thread Safe: concurrent program yields expected results (esp. regarding order)

Nondeterministic Execution: Sometimes one thread finished before the other and prints out its result in a seemingly wrong order

Amdahl vs. Gustafson

Amdahl:

fixed **relative** sequential portion

Gustafson:

fixed **absolute** sequential part

Amdahl & Gustafson

Computing Work	W
Sequential Execution Time	T_1
Number of CPUs	p
Performance loss	$\frac{T_1}{p} > \frac{T_1}{p}$ (reality)
Perfect	$T_p = \frac{T_1}{p}$
Magic	$T_p < \frac{T_1}{p}$
Parallelizable part	W_p
Sequential part	W_s
Speedup	S_p
Efficiency	$E_p = \frac{S_p}{p}$
Non-parallelisable fraction	λ

non-parallelisable fraction λ

$$W_s = \lambda \cdot W$$

$$W_p = (1 - \lambda) \cdot W$$

$$T_1 = W_s + W_p$$

$$T_p \geq W_s + \frac{W_p}{p}$$

Amdahl's Law

$$S_p = \frac{T_1}{T_p} \leq \frac{W_s + W_p}{W_s + \frac{W_p}{p}} = \frac{1}{\lambda + \frac{1-\lambda}{p}}$$

$$\Rightarrow S_\infty \leq \frac{1}{\lambda}$$

Gustafson's Law

Work that can be executed by 1 processor in time T : $W_s + W_p = T$

Work that can be executed by p processors in time T :

$$W_s + p \cdot W_p = \lambda \cdot T \quad p \cdot (1 - \lambda) \cdot T$$

$$S_p = \frac{W_s + p \cdot W_p}{W_s + W_p} = p \cdot (1 - \lambda) + \lambda = \\ = p - \lambda(p - 1)$$

Performance Model

T_∞ : Critical path, execution time on ∞ processors. Longest path from root to sink. **exam relevant:** sum the node values for actual "length" of the path. Longest path gives you T_∞ .

$\frac{T_1}{T_\infty}$: Parallelism. Wider = better.

Work Law: $T_p \geq \frac{T_1}{p}$

Span Law: $T_p \geq T_\infty$

Greedy Scheduler

"Teilt zu jeder Zeit so viele Tasks zu Prozessoren wie möglich"

Theorem 45: On an ideal parallel computer with p processors, a greedy scheduler executes a multi-threaded computation with work T_1 and span T_∞ in time:

$$T_p \leq \frac{T_1}{p} + T_\infty$$

Race Conditions

Race Conditions occur, when the result of a computation depends on scheduling.

Data race

Fehlerhaftes Programmverhalten verursacht durch ungenügend synchronisierten Zugriff zu einer gemeinsam genutzten Ressource, z.B. gleichzeitiges Lesen/Schreiben oder Schreiben/Schreiben zum gleichen Speicherbereich.

Bad interleaving:

Fehlerhaftes Programmverhalten verursacht durch eine unglückliche Ausführungsreihenfolge eines Algorithmus mit mehreren Threads, selbst dann wenn die gemeinsam genutzten Ressourcen anderweitig gut synchronisiert sind.

Rule of thumb:

Compiler and hardware are allowed to make changes that do not effect the semantics of a sequentially executed program. They insbesondere don't guarantee

that global ordering of memory access is provided as in the sourcecode (= fucks up the order of commands). C++ provides guarantees when synchronisation with a mutex is used.

Basically guarantees that reading and writing happens in one step, leaving no time for race conditions.

```
template <atomic>
std::atomic_int x{0};
std::atomic_int y{0};
```

Advanced C++

Memory Allocation

Rule of Five

If one of them is tinkered with, all of them should be:

- destructor
- copy-constructor
- copy-assignment operator
- move constructor
- move assignment operator

Copy-&-Swap Idiom

all members of `*this` are exchanged with members of `cpy`. When leaving `operator=`, `cpy` is cleaned up (deconstructed), while the copy of the data of `v` stay in `*this`.

```
class Vector{
    // move constructor
    Vector (Vector&& v) : Vector() {
        swap(v);
    }
    // normal assignment
    Vector& operator= (const Vector&
                        v){
        Vector cpy;
        swap(cpy);
        return *this;
    }
    // move assignment
    Vector (Vector&& v){
        swap(v);
        return *this;
    }
private:
    void swap(Vector& v){
        std::swap(sz, v.sz);
        std::swap(elem, v.elem);
    };
};
```

Templates

```
template <typename T> \\ templated
function
void swap(T& x, T&y){
    T temp = x; x = y;
    y = temp;
}
```

```
template <typename T> \\ templated
printer
void output(const T& t){
    for (auto x: t) \\ ranged for
        loop
        std::cout << x << " ";
    std::cout << "\n";
}
```

Iterators

```
// for support of ranged for-loops
class Vector{

    // Iterator (const in front for
    const_it)
    double* begin(){
        return elem;
    }

    double* end(){
        return elem+sz;
    }
};
```

Dynamic Memory Guidelines

For each `new`; a matching `delete!`
Avoid:

- **Memory Leaks:** Old object not freed/hog memory
- **Dangling Pointers:** Pointer to released object
- **Double Free:** Releasing memory more than once using `delete`

- i.e. use Smart Pointers!

Smart Pointers

```
std::unique_ptr<Node> nodeU(new
    Node());
std::shared_ptr<Node> nodeS(new
    Node());
```

- Never call `delete` on a pointer contained in a smart pointer
- Where possible, use `std::unique_ptr`
- If using `std::shared_ptr` make sure there are no cycles in the pointer graph.
- Avoid `new`, instead:

```
std::unique_ptr<Node> nodeU =
    std::make_unique<Node>()
std::shared_ptr<Node> nodeS =
    std::make_shared<Node>()
```

Lambdas & Functors

[value]	(int x)	->bool	{return x > value;}
capture parameters	return type	statement	

Functors are just classes with an overloaded `operator()`, so that when you use that operator it looks like you're calling a function (which you do, more specifically a member function). Functors can store member variables in between calls ("states").

```
// this is a functor
class LessThan{
    T value;
public:
    LessThan(T x) : value(x) {} // constructor
    bool operator() (T otherval) const{
```

```
        return (otherval < value);
    }

    // this is a lambda expression
    auto f = [value](int x){return x <
        value;};
    // [capture](parameters)->
    optional_returntype{the actual
        code;};
}

// things that go inside a capture:
// [x]: pass x by value
// [&x]: pass x by reference
// [&]: pass everything by ref.
// [=]: pass everything by value
// [&x,y]:pass x by ref. and y by
value
```

This cheat sheet is to be used as inspiration and not as a ready-made cheat sheet for the exam!

Snippets

Bridge

```

class Bridge {
private:
    unsigned int capacity = 0; // assert(capacity <= 3)

    std::mutex m;
    std::unique_lock<std::mutex> guard = std::unique_lock<std::mutex>(m);
    std::condition_variable cond;

public:
    // make sure that only three cars or one truck can enter!
    void enter_car(){
        std::cout << "Car trying to enter the bridge" << std::endl;

        guard g(m); // lock m
        cond.wait(g, [&]{return (capacity < 3);}); // wait until cap < 3

        std::cout << "Car ENTERING the bridge" << std::endl;

        ++capacity; // car is now on bridge
    }

    // make sure that waiting cars or even trucks can enter
    void leave_car(){
        guard g(m); // lock m
        --capacity; // car is now on bridge
        cond.notify_all(); // tell waiting vehicles
                           // they can (maybe) go
        std::cout << "Car LEAVING the bridge. capacity = "
                  << capacity << std::endl;
    }

    // make sure that only three cars or one truck can enter!
    void enter_truck(){

```

```

        std::cout << "Truck trying to enter the bridge"
                  << std::endl;

        guard g(m); // lock m
        cond.wait(g, [&]{return (capacity == 0);}); // wait until cap == 0

        std::cout << "Truck ENTERING the bridge" << std::endl;

        capacity += 3; // truck is now on bridge
    }

    // make sure that waiting cars or even trucks can enter
    void leave_truck(){
        guard g(m); // lock m
        capacity -= 3; // car is now on bridge
        cond.notify_all(); // tell waiting vehicles
                           // they can (maybe) go
        std::cout << "Truck LEAVING the bridge. capacity = "
                  << capacity << std::endl;
    }
}

```

Online store

```

class Item {
private:
    int rating_sum = 0;
    int rating_count = 0;
    std::recursive_mutex mtx;
public:
    Item() {};
    /* Returns average rating. 0 if no rating occurred */
    double get_rating() {
        std::lock_guard<std::recursive_mutex> lock(mtx);
        if(rating_count == 0) return 0.0;
        return (double)rating_sum / rating_count;
    }
    void add_rating(int stars){
        assert(1 <= stars && stars <= 5);

```

```

        std::lock_guard<std::recursive_mutex> lock(mtx);
        rating_sum += stars;
        rating_count++;
    }
    void out_rating(){
        std::lock_guard<std::recursive_mutex> lock(mtx);
        std::cout << "ratings:" << rating_count << "\n";
        std::cout << "score:" << get_rating() << "\n";
    }
}

```

Levenshtein Distance DP

```

#include <string>
#include <vector>

void mprint(int** M, unsigned int xlen, unsigned int ylen){
    for(unsigned int i = 0; i <= xlen; i++){
        std::cout << std::endl;
        for(unsigned int j = 0; j <= ylen; j++){
            std::cout << M[i][j] << " ";
        }
    }
}

void print(std::string s){
    std::cout << s << std::endl;
}

unsigned int min(unsigned int v, unsigned int w){
    return (v <= w ? v : w);
}

// action cost
// Ins(a) = 1;
// Del(a) = 1;
// Rpl(a, e) = 1;
// Rpl(a, a) = 0;

```

! This cheat sheet is to be used as inspiration and not as a ready-made cheat sheet for the exam!

```

// returns Levenshtein distance of two given
// strings
unsigned Levenshtein(const std::string& x, const
    std::string& y){
    const unsigned int xlen = x.length();
    const unsigned int ylen = y.length();

    // std::vector<std::vector<unsigned int>> C;
    unsigned int** C = new unsigned int*[xlen+1];
    for (unsigned int i = 0; i <= xlen; ++i){
        C[i] = new unsigned int[ylen+1];
    }

    // fill the edge cases in 0th row
    for(unsigned int j = 0; j <= ylen; j++){
        C[0][j] = j;
    }
    //      - P I Z Z A
    //      C 0_1_2_3_4_5
    //      - 0|0 1 2 3 4 5
    //      F 1|x x x x x x
    //      R 2|x x x x x x
    //      Y 3|x x x x x x
    for(unsigned int i = 1; i <= xlen; i++){
        C[i][0] = i; // fill egde cases in 0th col
        //      - P I Z Z A
        //      C 0_1_2_3_4_5
        //      - 0|0 1 2 3 4 5
        //      F 1|1 x x x x x
        //      R 2|2 x x x x x
        //      Y 3|3 x x x x x
        for(unsigned int j = 1; j <= ylen; j++){
            if(x[i-1] == y[j-1]){
                C[i][j] = C[i-1][j-1]; // case: Replace(a,
                    a)
            }else{
                unsigned int m;
                m = min(C[i-1][j] + 1 // case: Insert(a)
                    , C[i-1][j]); // case: Delete(a)
                m = min(C[i-1][j-1] + 1, m); // case:
                    Replace(a
                    , e)
                C[i][j] = m;
            }
        }
    }
}

```

! This cheat sheet is to be used as inspiration and not as a ready-made cheat sheet for the exam!

```

        return C[xlen][ylen];
    }

DFS Adjacency Matrix

using Node = std::size_t;

// color for graph traversal
enum class Color {
    white, grey, black
};

using Colors = std::vector<Color>;

// DFS on graph <g> starting from node <v> with
// coloring nodes (grey coloring not used here)
void DFS(const DiGraph& g, Node v, Colors& colors){
    colors[v] = Color::grey;
    for (size_t i = 0; i < g.number_nodes(); ++i){
        if (g.getEdge(v, i) > 0){
            if (colors[i] == Color::white){
                DFS(g, i, colors);
            }
        }
    }
    colors[v] = Color::black;
}

// DFS on graph g starting from node v, uses
// grey-coloring to detect cycles
// returns if cycle can be found starting from
// node v
bool findCycle(const DiGraph& g, Node v, Colors&
    colors){
    colors[v] = Color::grey;
    for (size_t i = 0; i < g.number_nodes(); ++i){
        if (g.getEdge(v, i) > 0){
            if (colors[i] == Color::white){
                if (findCycle(g, i, colors)) return true;
            } else if (colors[i] == Color::grey){
                return true;
            }
        }
    }
}

```

```

    colors[v] = Color::black;
    return false;
}

// post: return if graph is symmetric
// i.e. edge is present iff its antiparallel
// edge is present
bool DiGraph::isSymmetric(){
    for (Node i = 0; i < n; ++i){
        for (Node j = i+1; j < n; ++j){
            if (getEdge(i,j) != getEdge(j,i)){
                return false;
            }
        }
    }
    return true;
}

// post: returns true if there is a path from node
// <from> to node <to>
// via the directed edges of the graph
bool DiGraph::hasPath(Node from, Node to){
    Colors colors(n, Color::white);
    DFS(*this, from, colors);
    return colors[to] != Color::white;
}

// post: returns true iff from node <from> a cycle
// can be reached
// by traversing forward along the directed
// edges of the graph
bool DiGraph::hasCycle(Node from){
    Colors colors(n, Color::white);
    return findCycle(*this, from, colors);
}

// post: returns true iff the graph contains no
// cycle
bool DiGraph::isAcyclic(){
    for (Node i = 0; i < n; ++i){
        if (hasCycle(i))
            return false;
    }
    return true;
}

```