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$f(n) = O(g(n))$ , if  $f(n)/g(n) \leq C$  or  $\lim_{n \rightarrow \infty} f'(n)/g'(n) = 0$  (limit asymptotic theorem)  
 $f(n) = \Omega(g(n))$ , if  $f(n)/g(n) > C$  or  $\lim_{n \rightarrow \infty} f'(n)/g'(n) = \infty$   
 $f(n) = \Theta(g(n))$ , if  $O(n) = \Omega(n)$  or  $\lim_{n \rightarrow \infty} f'(n)/g'(n) = 0$

TODO: LOG-RULES TODO: arithmetic/geometric sums

TODO: data-structures and algorithms time complexity Topological (vertices connected left to right):  $O(|V| + |E|)$  Dijkstra (shortest path to all vertices):  $O(|V|^2)$  Prim's (minimum spanning tree):  $O(|V|^2)$  BFS (find all vertices accessible):  $O(|V| + |E|)$  Heap: SelfBalancingTree: EdmondsKarp:  $O(|E| |f|)$ ,

## Divide and Conquer

$T(n) = a \cdot T(n/b) + f(n)$ ,  $a = \text{num-subproblems}$ ,  $b = \text{size-subproblems}$  critical:  
 $n^{\log_b a}$ ,  $< (n^{\log_b a})$ ,  $= (n^{\log_b a} \cdot \log_2)$ ,  $(> \text{ and } a \cdot f(n/b) \leq c \cdot f(n))$   $(f(n))$  **Binary-Search** If possible for  $n$ , possible larger values; monotonicity. Also have upper/lower bound

**Algorithm 1.** Recursively divide into two subarrays of approximately equal parts. Find distinct cards in the first  $k/2$  and last  $k/2$ . 2. Merge the results of 2 subarrays by checking through each card one by one in both halves to get a subarray with distinct cards. 3. Base case: for  $n=1$  students, a single collection contains distinct cards **Induction** For base case of  $n=1$ , we know a single collection has distinct cards. Assume this works for  $n=k$ . For  $n=k+1$ , by problem definition, merging two collections will always produce a collection of distinct cards. So, merging two collections of  $n=k$  will produce a distinct collection at  $n=k+1$ . ##### Greedy **Stays-Ahead 1.** Let greedy solution be  $G=(g_1, g_2, g_3, \dots, g_n)$  where  $g_i$  represents a particular rod. Let an alternative supposed optimal solution be  $O=(o_1, o_2, o_3, \dots, o_n)$  2. Base case is welding the 2 shortest rods. Welding  $g_1 + g_2$  will yield the absolute shortest welded rod of absolute minimal cost of any 2 rods. Therefore, welding  $g_1 + g_2$  costs no more than welding  $o_1 + o_2$  3. Assume that welding rods up to  $g_{k-1}$  costs no more than welding rods up to  $o_{k-1}$

4. As the cost of the resultant rod from  $g_{k-1}$  welds is no more than  $o_{k-1}$ ,  $g_{k-1} + g_k$  cannot cost any more than  $o_{k-1} + o_k$ . As a result, since  $O$  is arbitrary,  $G$  must

be optimal *ALTERNATE: If these rods were not present at this location in  $O$ , then they must appear closer to the centre of  $O$ .* **Exchange** 1. Let  $x$  be the activity that starts last overall 2. Consider an alternative schedule  $S$  that doesn't contain  $x$  3. Let  $y$  be that activity in  $S$  that starts last 4. Since we know  $x$  is activity that starts last,  $x$  must start after  $y$ .  $y$  won't clash, meaning  $x$  won't clash 5. Therefore, can iteratively transform  $S$  into a new schedule  $S'$  that contains the activity that starts last **TODO: Contradiction** Moreover, since the greedy algorithm deletes vertices from  $G$ , it must have arrived at graphs that contain  $H$  as a subgraph. In particular, the greedy algorithm must arrive at  $H'$  before  $H$  Suppose that there exists an alternative solution that deletes a smaller set of vertices  $D = (v_0, v_1, \dots)$  at this point than our solution. Therefore, there must exist at least one vertex  $v_i$  in our solution not in  $D$ . For  $v_i$  to not have been deleted, it must not be adjacent to at least  $k$  vertices. However, our algorithm only deletes vertices that are not adjacent to less than  $k$  vertices, i.e. wouldn't delete  $v_i$ . This contradicts assumption that  $D$  represents a smaller optimal solution. Therefore, algorithm is optimal.

## Flow

Create a flow network with: - Source vertex  $S$  and sink vertex  $T$  -  $n$  children vertices  $R$  - For each child  $i$ , an edge  $(S, R)$  with capacity 1 **Correctness** For particular edge, flow conservation ensures that cannot receive more ... For particular vertex, capacity constraint ensures cannot hold more ... Maximum flow is constrained by the total number of children, i.e.  $|f| = n$

## Dynamic Programming

**Subproblem:** Let  $\min(i, a)$  be minimum time taken to travel to city  $c_i$  arriving on animal  $a$ . Let  $\text{animal}(i)$  be the animal that was used to arrive at city  $c_i$  on a journey of minimal time. **Recurrences:**  $\min(i, a) = \text{minimum}\{\min(i-1, a_1) + d(i)/v(a), \text{ if } R(i-1, a_1, a)\}$   
 $\text{animal}(i) = \text{argmin}\{\min(i, a)\} a \rightarrow G, M, A, I, L$  **Base Case: Order:** Initialise 2D table  $\min[n][5]$  bottom-up with an outer loop from  $i = 1..n$  and inner loop  $j = 1..5$  Set all entries to  $\infty$  to handle cases where no animal can arrive at a particular city. To obtain list of animals, backtrack from  $i = n..1$  **Final**

**Solution:** Minimal Amount of Time:  $\text{minimum}\{\min(n, a)\}$  **Proof:** 1. Base case  
 2. Assume that optimal determined by the optimal choice of between consecutive cities  $c_k$  and  $c_{k+1}$ . Suppose an alternative optimal solution is where the choice between consecutive cities is not optimal. In this case, the total travel time could be reduced by changing the animals between these cities. For example, if this solution has animals  $a_1, a_2$  there exists animals  $a_3, a_4$  such that smaller. This contradicts assumption that this alternative solution is optimal. Therefore, optimal solution is built up from optimal solutions to its subproblems. 3. Explaining recurrence, considering minimum of all allowable combinations between  $c_{i-1}, c_i$  the optimal choice for  $c_{i-1}, c_i$  will be made.

## Strings

## Linear

**maximise**  $P = 5x_1 + 3x_2 + 4x_3$   $x_1 + 2x_2 + x_3 \leq 6$   
 $3x_1 + x_2 + x_3 \leq 4$  **minimise**  $P^* = 6y_1 + 4y_2$   
 $y_1 + 3y_2 \geq 5$   
 $2y_1 + y_2 \geq 3$   
 $y_1 + y_2 \geq 4$  unconstrained  $x \rightarrow \mathbb{R}; x = x' - x''$

## Intractability

algorithm terminates correctness time-complexity

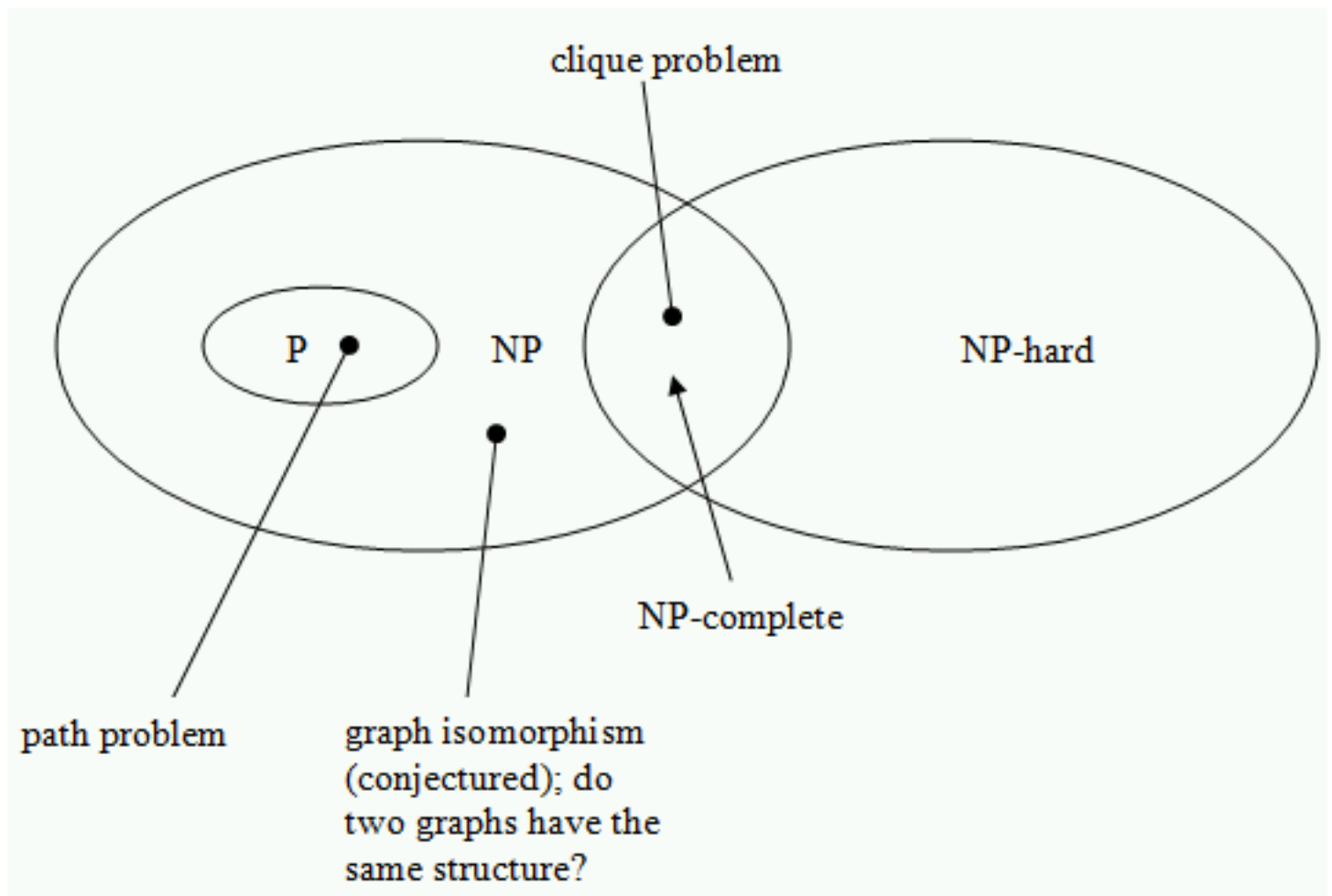


Figure 1: Venn-Diagram