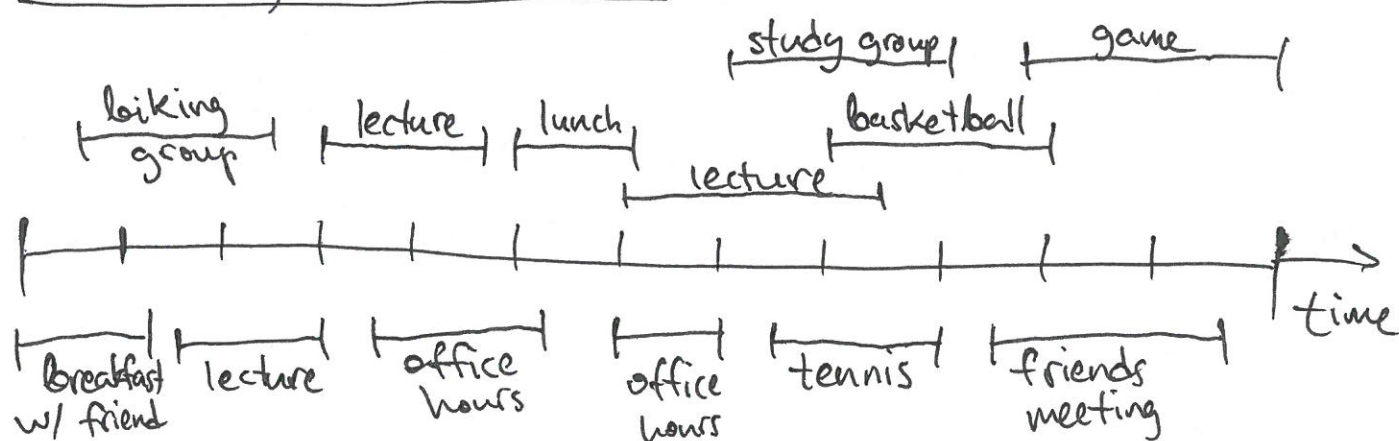


# Planning Your Time

(1)



Input Activities,  $i=1, \dots, n$ , each has a start time  $s_i$  & a finish time  $f_i$ .

A schedule Subset of <sup>non-overlapping</sup> activities  $i_1, \dots, i_k$   
 $s_{i_1} < f_{i_1} \leq s_{i_2} < f_{i_2} \leq \dots < f_{i_k}$

(A) "I want to do as many activities as possible!" (i.e., max  $k$ ).



People often use greedy strategies:

pick an activity that looks "best" now, then "deal with the consequences" move on to next. E.g.

\* Pick shortest activity (i.e., min  $f_i - s_i$ )

\* Pick most urgent activity (i.e., min  $s_i$ )

\* Pick earliest ~~deadline~~ <sup>finish</sup> (i.e., min  $f_i$ )

(2)

Advantage Easy to implement, efficient to find next activity.

Run-time for the three algorithms we suggested:  $O(n \lg n)$

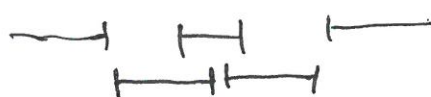
1 sort the activities <sup>(i)</sup> (according to  $f_i - s_i$  or  $s_i$ )  
 (ii) sort  $s_1 \dots s_n, f_1 \dots f_n$

2 pick min activity, rule out overlapping activities  
 (i.e., start time between start & finish of activity)

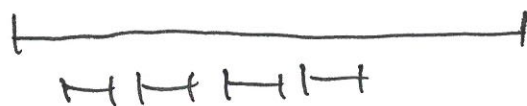
Step 1 takes  $O(n \lg n)$  time. Step 2 takes  $O(n)$  time overall. ✱

Disadvantage May not find optimal solution.

\* Shortest activity - one short activity may rule out two activities



\* Most urgent activity - urgent activity may rule out many activities that start later

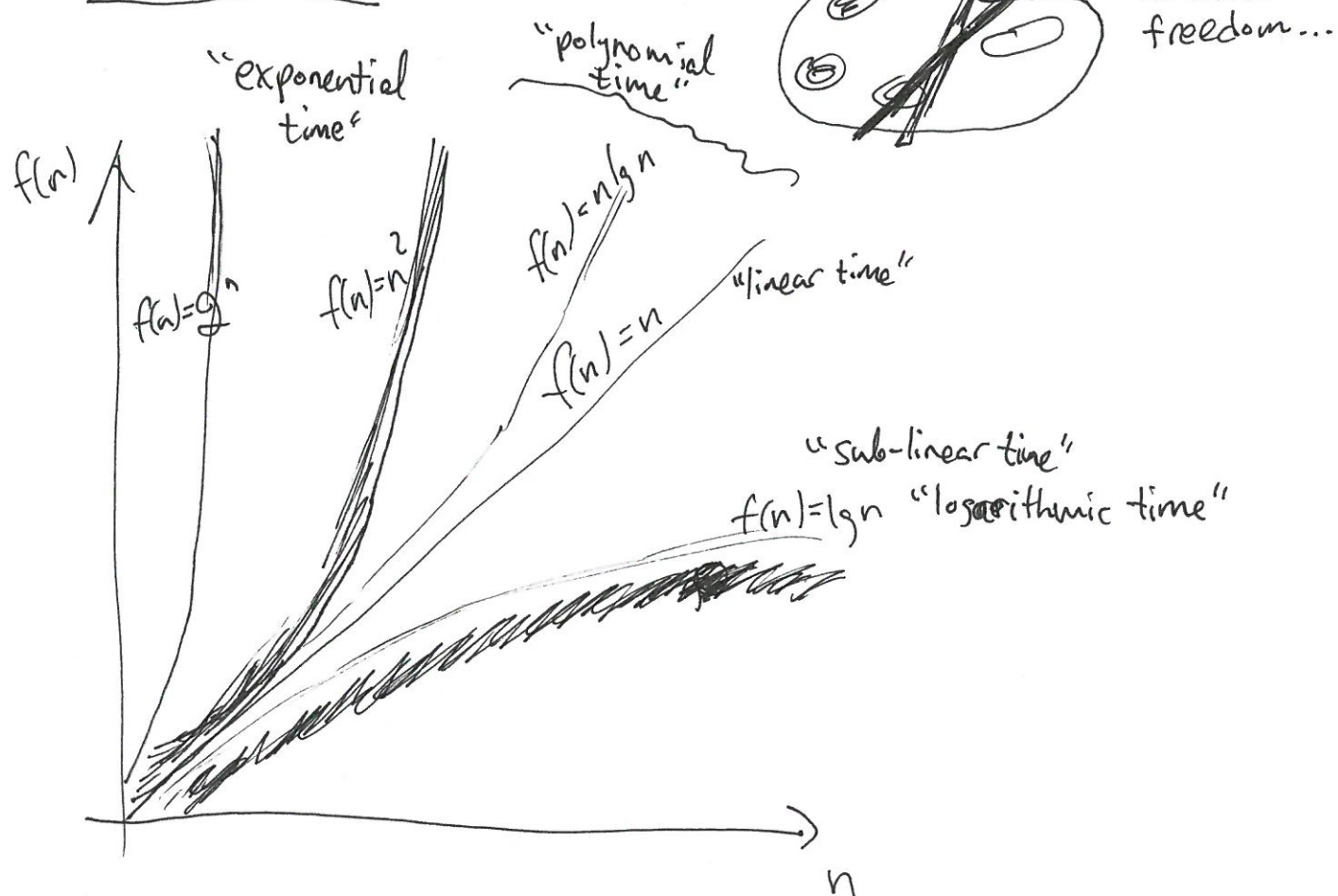


\* Earliest finish time - works for previous examples, and in fact works in general as we'll see later in the term.

Intuition: put yourself ahead of the game for the next activities.



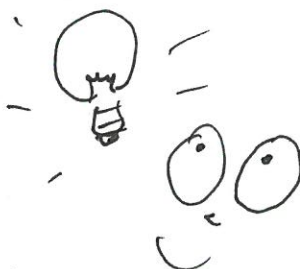
# \* Run-times



\* We use "O" notation:  $f(n) = O(g(n))$  if  $\exists m, n_0 \forall n \geq n_0$   $f(n) \leq m \cdot g(n)$

Rationale: one operation can count as two or more in a different computational model. We want to count # operations while ignoring such differences.

(B) "I have priorities"



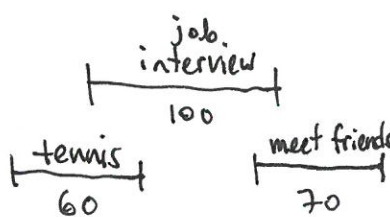
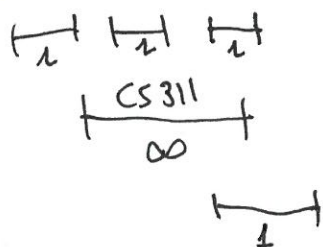
(4)

Input Activities as before but now every activity has a weight  $w_i$

Goal  $\max \sum_{j=1}^k w_{i_j}$  for a schedule  $i_1 \dots i_k$ .

Note The no-weights case is a special case with  $w_i = 1 \forall i$ .

Examples

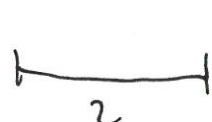


Observation The greedy approach fails!

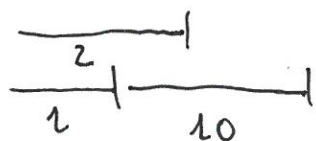
What's better ??

 get less, free fast

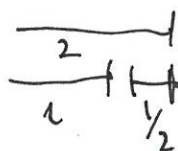
OR

 get more, free later

Depends... Are there exciting opportunities ahead?



OR

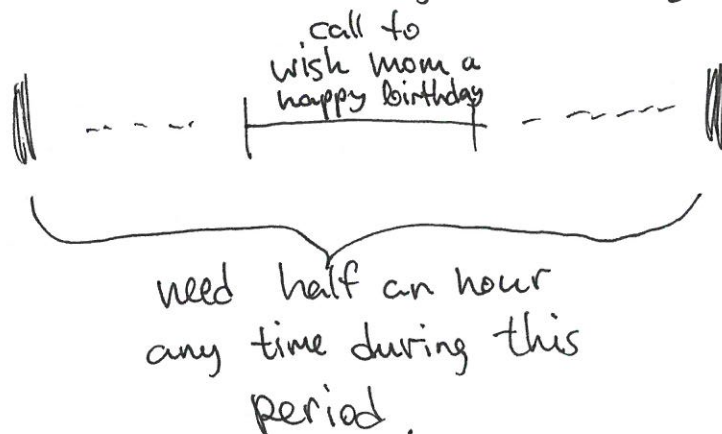




Later in the semester we'll discuss "dynamic programming", a powerful algorithmic paradigm that records information about different alternatives.

Note There can be up to  $2^n$  different schedules. Going through them one by one - <sup>using</sup> "brute force" - would be extraordinarily inefficient. In contrast - Dynamic programming gives an  $O(n \log n)$  time algo for the problem!!

(C) "I'm flexible in my scheduling"



NP-hard

The most efficient algo we know takes exponential time!!

Input  $n$  activities; each has a start time  $s_i$ , a finish time  $f_i$  and a duration  $l_i$ .

Goal Maximize  $k$  with  $i_1 \sim i_k$  & s.t.  $s_{i_j} \leq s_{i_{j+1}} < f_{i_j} \leq f_{i_{j+1}}$   
 $s_{i_1} < f_{i_1} < \dots < s_{i_k} < f_{i_k}$