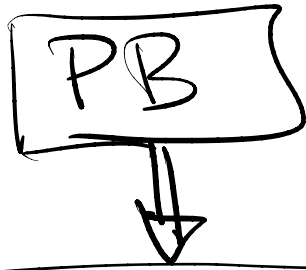


Divide & Conquer

DP = combination of subpb-solutions.

↓
split into subproblems

Greedy Algorithms



Greedy

- Divide (Split)
- Decide
- Break

SUBPB

- Solve SubPb

- [Sol] = combine (SubPb, Sol)

Dynamic Programming

- look at all possible SUBPB
dont know how to break it

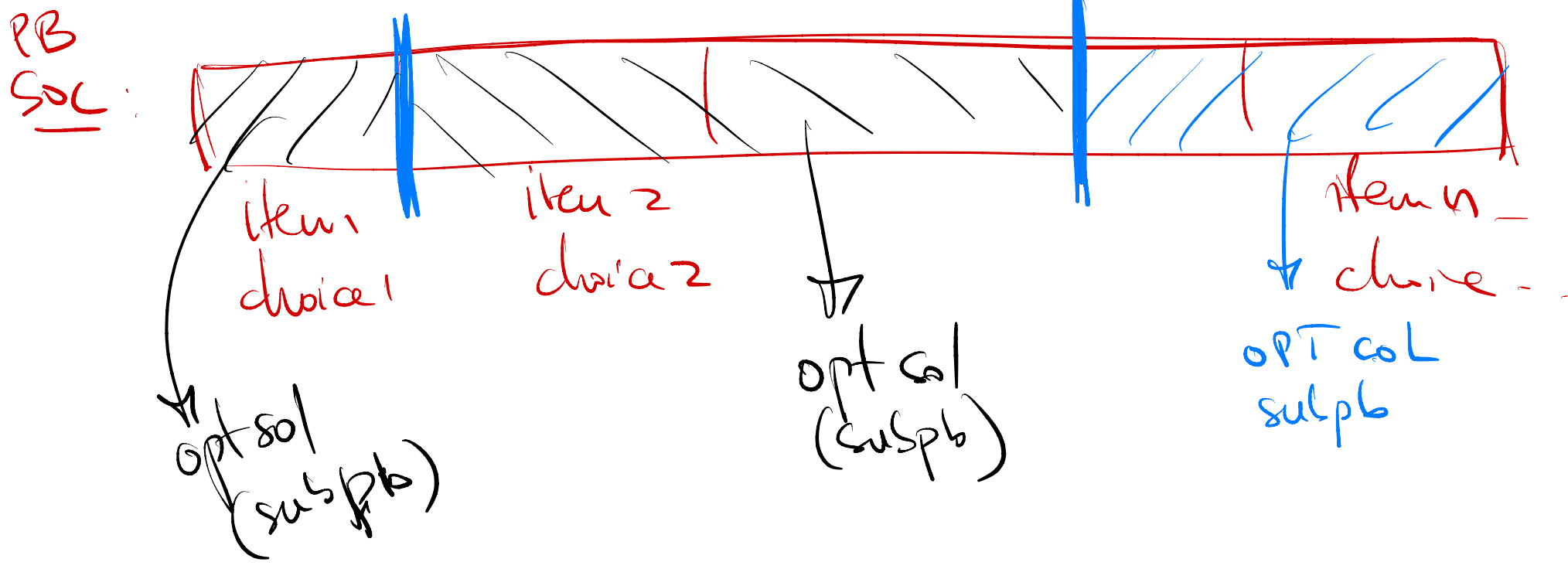
- solve all possible SUBPB
(even ones we dont need)

- given SOL (subpb) decide the split \Rightarrow which SUBPB we need

- [SOL] = comb (selected subpb, SOL)



Sol \rightarrow Subpb_sol Optimal Structure



Objective

$$C[\text{input}_{PB}] = \text{objective of the pb} = \text{value (split/decision)} + \text{combine function } \left\{ C[\text{input}_{subpb}] \right\}$$

Week 5 Objectives

- Subproblem structure
- Greedy algorithm
- Mathematical induction application
- Greedy correctness

Subproblem Optimal Structure

- Divide and conquer – optimal subproblems
- divide PROBLEM into SUBPROBLEMS, solve SUBPROBLEMS
- combine results (conquer)
- **critical/optimal structure**: solution to the PROBLEM must include solutions to subproblems (or subproblem solutions must be combinable into the overall solution)
- PROBLEM = {DECISION/MERGING + SUBPROBLEMS}

Optimal Structure - GREEDY

- PROBLEM = {DECISION/MERGING + SUBPROBLEMS}
- GREEDY CHOICE: can make the DECISION without solving the SUBPROBLEMS
 - the GREEDY CHOICE looks good at the moment, and it is globally correct
 - example : pick the smallest value
 - solve SUBPROBLEMS after decision is made
- GREEDY CHOICE: after making the DECISION, very few SUBPROBLEMS to solve (typically one)

Optimal Structure - NON GREEDY

- *dynamic* Cannot make a choice decision/CHOICE without solving subproblems first
- Might have to solve many subproblems before deciding which results to merge.

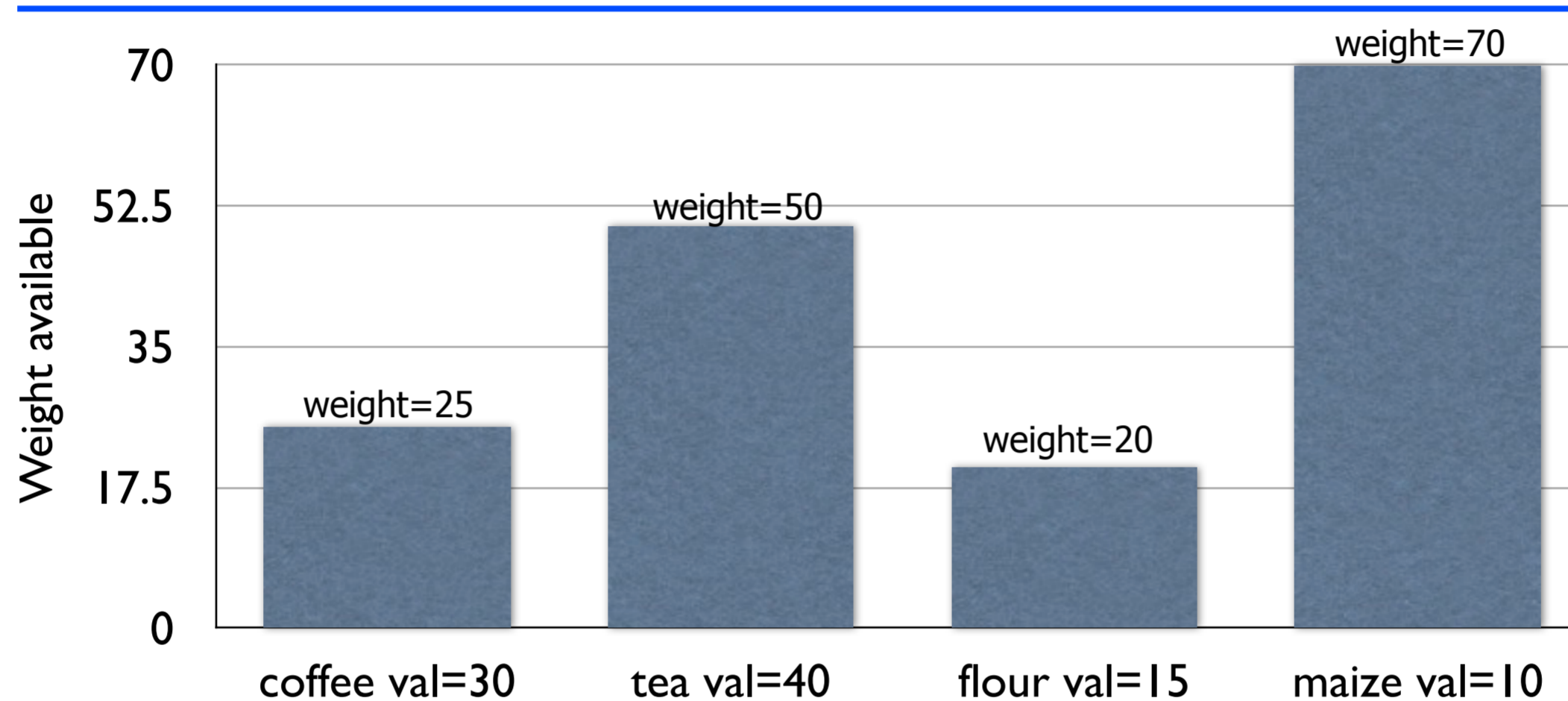
Ex: ~~Fractional~~ Knapsack

Greedy ← *discrete => DP*

discrete = paintings in museum } take it
leave it

- fractional goods (coffee, tea, flour, maize...) sold by weight
- supply (weights/quantities available) $w_1, w_2, w_3, w_4, \dots$
- values (totals) $v_1, v_2, v_3, v_4, \dots$
 - ex: coffee $w_1=10$ pounds; coffee overall value $v_1=\$40$
- knapsack capacity (weight) = W *weight constraint*
- task : fill the knapsack to maximize value

Ex: Fractional Knapsack

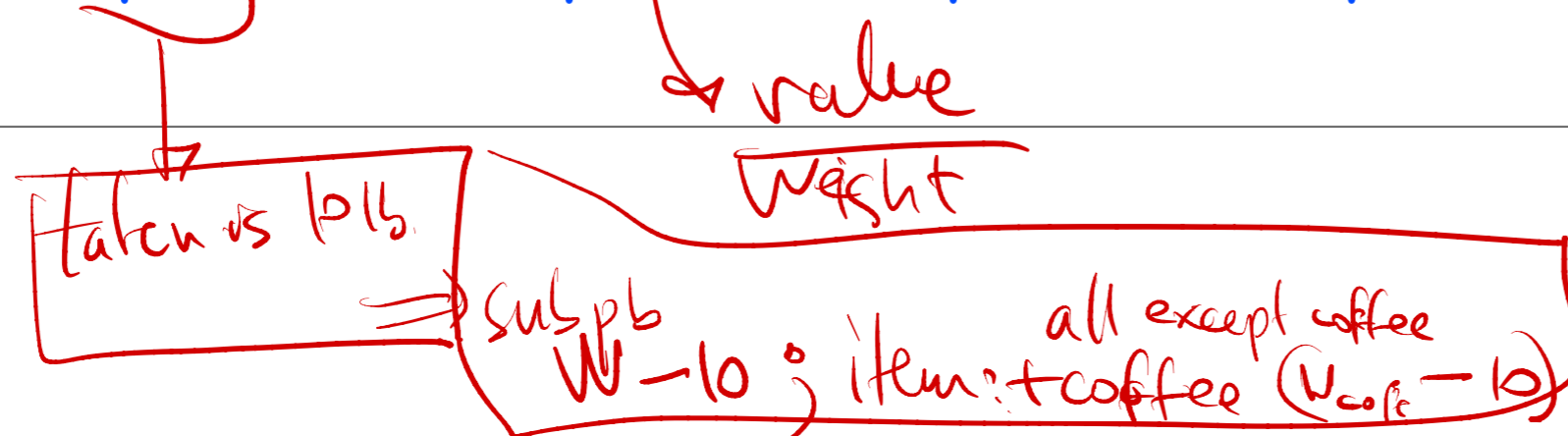


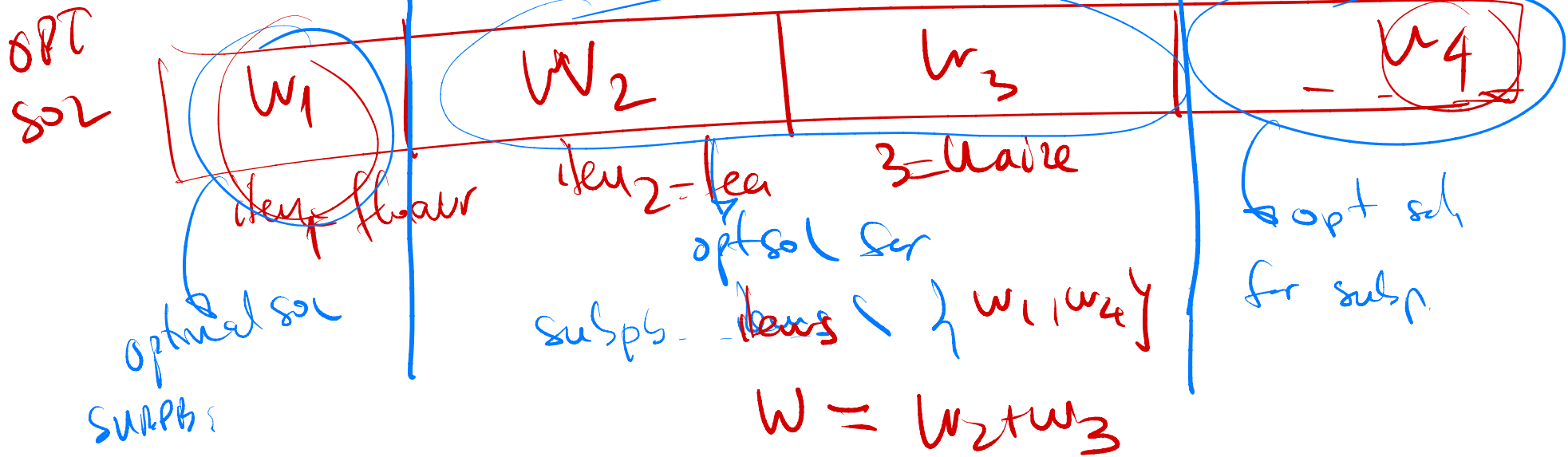
● naive approaches may lead to a bad solution

- choose by biggest value - tea first
- choose by smallest quantity - flour first

● choose by quality is correct - ^{Greedy} coffee first

- $q_{\text{coffee}}=30/25$; $q_{\text{tea}}=40/50$; $q_{\text{flour}}=15/20$; $q_{\text{maize}}=10/70$





$$C[W, \text{items}] = \left(V_{\text{best}} \left\{ \begin{array}{l} \text{arg max} \\ \text{max} \end{array} \frac{v}{w} \right\} \right)$$

W_{best} weight

$$\rightarrow C \left[\begin{array}{l} W - w_{\text{best}} \\ \text{items} - \end{array} \right]$$

w_{best} (circled in blue)

• greedy choice criteria

• greedy proof

Ex: Fractional Knapsack

- solution: compute item quality (value/weight)
- $q_i = v_i / w_i$
- sort items by quality $q_1 > q_2 > q_3 > \dots$
- LOOP
 - take as much as possible of the best quality
 - if knapsack full, STOP
 - if stock depletes (knapsack not full), move on to the next quality item, repeat
 - END LOOP

Proofs

• exchange
global excha sol \Rightarrow better value

• induction
after k Greedy steps \Rightarrow part of OPT SOL

• stay ahead,
after k Greedy steps
I have the same choices
(or better) as other solutions

Fractional Knapsack – greedy proof

- proving now that the greedy choice is optimal
 - meaning that the solution includes the greedy choice.
- greedy choice: take as much as possible from best quality (below item with quality q_1)
 - items available sorted by quality: $q_1 > q_2 > q_3 > \dots$, greedy choice is to take as much as possible of item 1, that is quantity w_1
- contradiction/exchange argument
 - suppose that best solution doesn't include the greedy choice: $SOL = (r_1, r_2, r_3, \dots)$ quantities chosen of these items, and that r_1 is not the max quantity available (of max quality item), $r_1 < w_1$
 - create a new solution SOL' from SOL by taking more of item 1 and less of the others
 - $e = \min(r_2, w_1 - r_1)$; $SOL' = (r_1 + e, r_2 - e, r_3, r_4, \dots)$
 - $\text{value}(SOL') - \text{value}(SOL) = e(q_1 - q_2) > 0$ which means SOL' is better than SOL : CONTRADICTING that SOL is best solution

Fractional Knapsack – greedy proof

- english explanation:

- say coffee is the highest quality,
- the greedy choice is to take max possible of coffee which is $w_1=10$ pounds

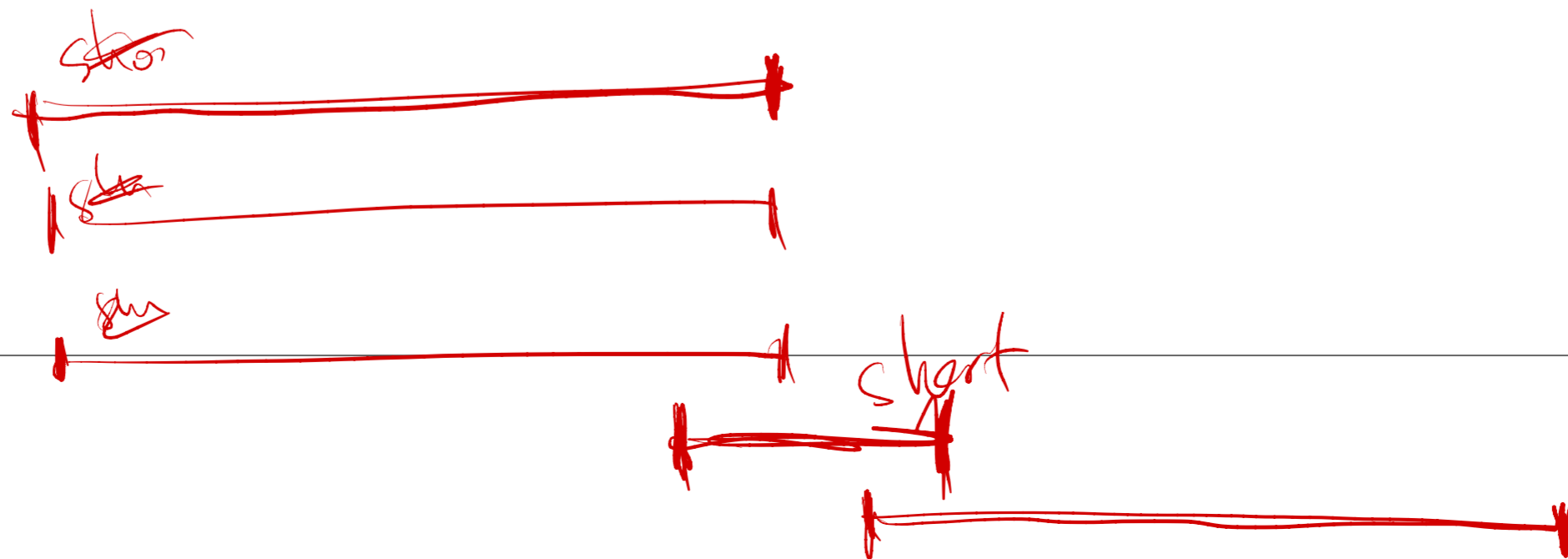
- contradiction/exchange argument

- suppose that best solution **doesn't include the greedy choice**:
 $SOL=(8\text{pounds coffee, } r_2 \text{ of tea, } r_3 \text{ flours,...})$ $r_1=8\text{pounds} < w_1=10\text{pounds}$
- create a new solution SOL' from SOL by taking out 2 pounds of tea and adding 2 pounds of coffee; $e=2\text{pounds}$
 - $e=\min(r_2, w_1-r_1)$; $SOL'=(r_1+e, r_2-e, r_3, r_4\dots)$
 - $\text{value}(SOL') - \text{value}(SOL) = e(q_1-q_2) > 0$ which means SOL' is better than SOL : **CONTRADICTING** that SOL is best solution

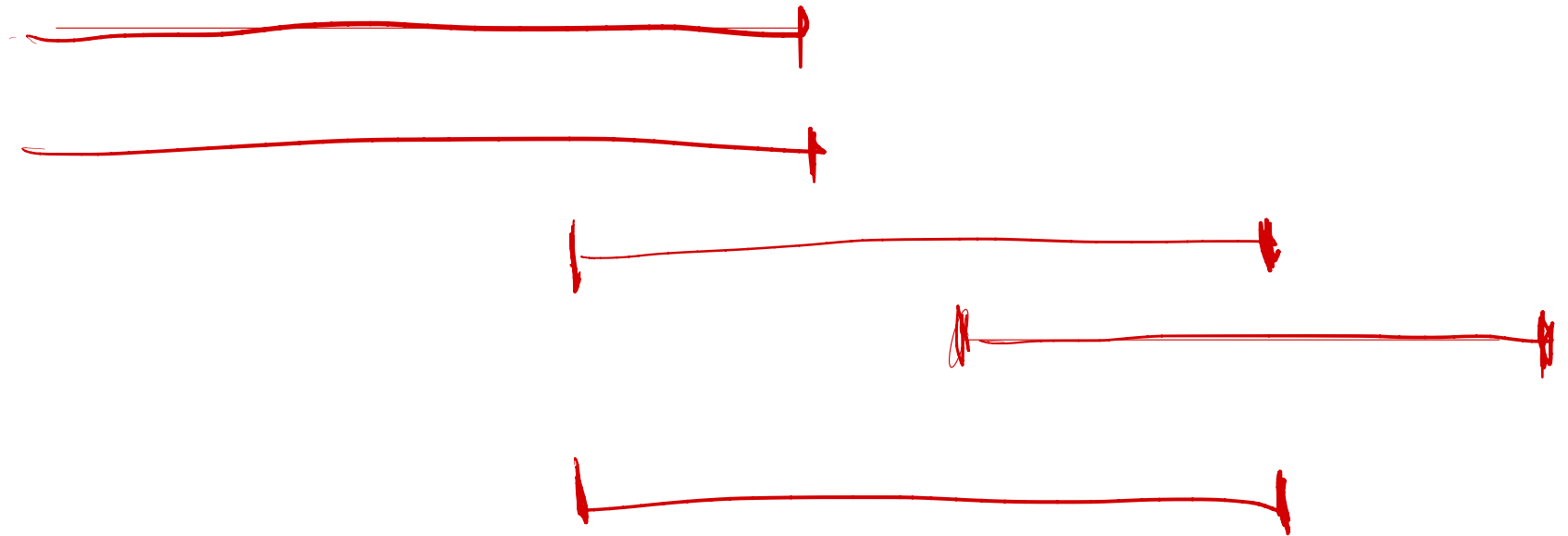
Activity Selection Problem

- S =set of n activities given by start and finish time
 $a_i = (s_i, f_i)$ $i=1:n$, $f_i > s_i$
- Determine a selection that gives a maximal set
 - select maximum number of activities
 - no overlapping activities can be selected

Criteria:
shortest
interval



criteria: least overlap (# of \cap activities)



Activity Selection Problem

- Greedy solution: sort activities by their finishing time
 - $f_1 < f_2 < f_3 \dots$
 - select the activity that finishes first $a = (s_1, f_1)$
 - discard all overlapping activities with selected one : discard all activities with starting time $s_i < f_1$
 - repeat
- intuition: activity that finishes first is the one that leaves as much time as possible for other activities

Activity Selection Problem

- Proof of greedy choice optimality
 - activities sorted by finishing time $f_1 < f_2 < f_3 \dots$
 - greedy choice pick the activity a with earliest finishing time f_1
 - want to show that activity a is included in one of the best solutions (could be more than one optimal selection of activities)
- Exchange argument
 - SOL a best solution.
 - if SOL includes a , done.
 - suppose the best solution does not select a , $SOL = (b, c, d, \dots)$ sorted by finishing time $f_b < f_c < f_d$. Then create a new solution that replaces b with a $SOL' = (a, c, d, \dots)$.
 - This solution SOL' is valid, a and c don't overlap: $s_c > f_b > f_a$
 - SOL' is as good as SOL (same number of activities) and includes a

Mathematical Induction

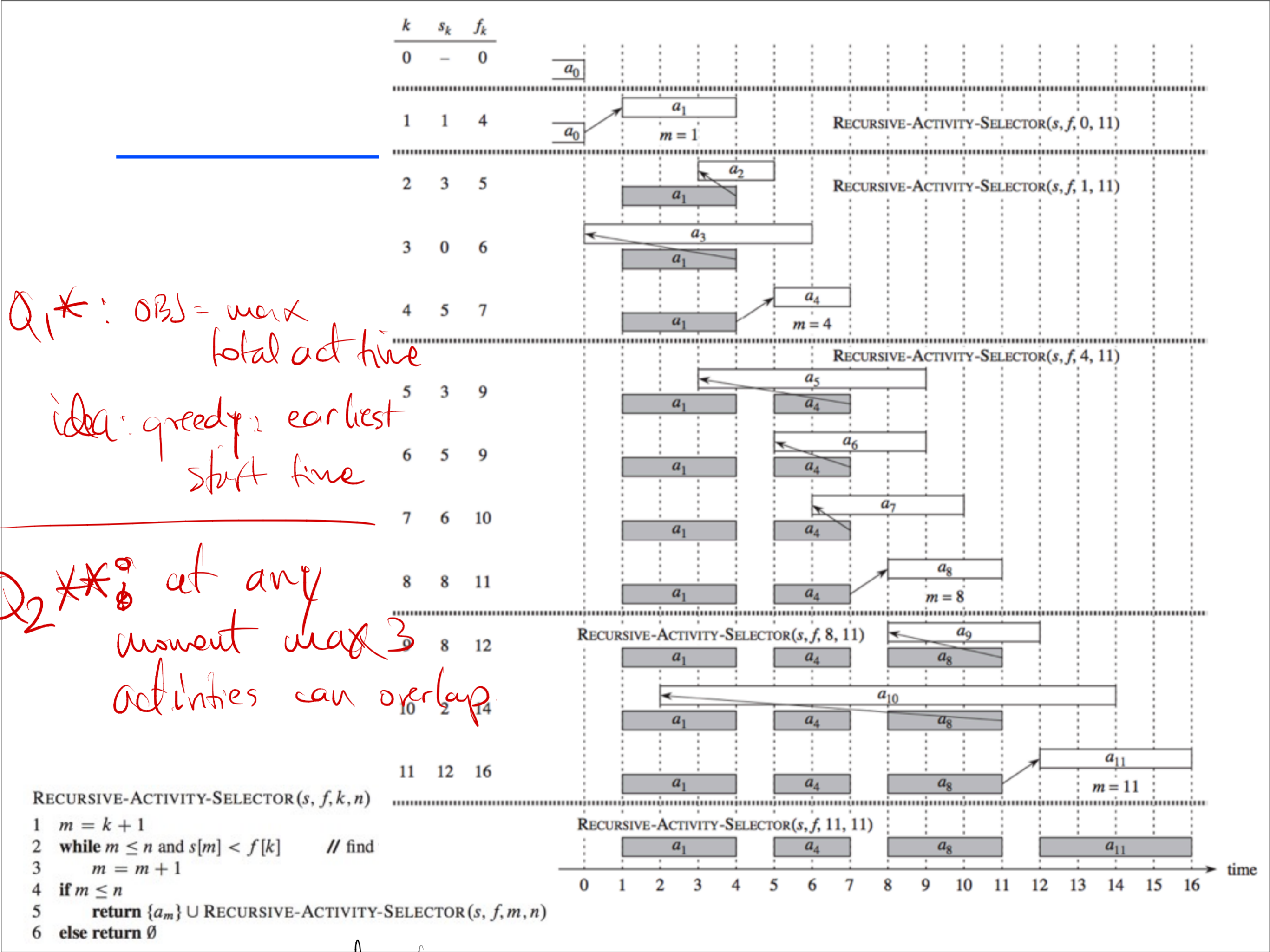
- property $P(n) = \{\text{TRUE}, \text{FALSE}\}$ for $n = \text{integer}$
 - want to prove $P(n) = \text{TRUE}$ for all n
- Base cases: $P(n) = \text{TRUE}$ for any $n \leq n_0$
- Induction Step: prove $P(n+1)$ for next value $n+1$
 - if $P(t) = \text{TRUE}$ for certain values of $t < n+1$ then prove by mathematical derivation/arguments that $P(n+1) = \text{TRUE}$
- Then $P(n) = \text{TRUE}$ for all n

Mathematical Induction- Example

- $P(n): 1+2+3+\dots+n = n(n+1)/2$
- base case $n=1 : 1=1*2/2$ - correct
- induction step : lets prove $P(n+1)$ assuming $P(n)$
 - $P(n+1) : 1+2+3+\dots+n + (n+1) = (n+1)(n+2)/2.$
 - assuming $P(n)$ TRUE : $1+2+3+\dots+(n+1) = [1+2+3+\dots+n] + (n+1) = n(n+1)/2 + (n+1) = (n+1)(n+2)/2$; so $P(n+1)$ TRUE
- thus $P(n)$ TRUE for all $n>0$

Activity Selection – Induction Argument

- $s(a)$ = start time; $f(a)$ = finish time
- $SOL = \{a_1, a_2, \dots, a_k\}$ greedy solution
 - chosen by earliest finishing time
- $OPT = \{b_1, b_2, \dots, b_m\}$ optimal solution, sorted by finishing time; optimal means m max possible
- prove by induction that $f(a_i) \leq f(b_i)$ for all $i=1:k$
 - base case $f(a_1) \leq f(b_1)$ because $f(a_1)$ smallest in the whole set
 - inductive step: assume $f(a_{n-1}) \leq f(b_{n-1})$. Then b_n is a valid choice for greedy at step n because $f(a_{n-1}) \leq f(b_{n-1}) \leq s(b_n)$. Since greedy picked a_n over b_n , it must be because a_n fits the greedy criteria $f(a_n) \leq f(b_n)$
- so $f(a_k) \leq f(b_k)$. If $m > k$ then any b_{k+1} item would also fit into greedy solution (CONTRADICTION) thus $m=k$



Q_1^* : OBJ = max total act time

idea: greedy: earliest start time

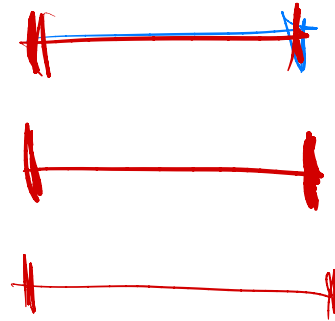
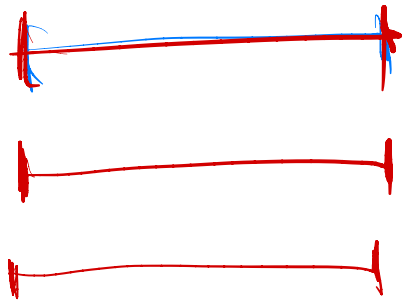
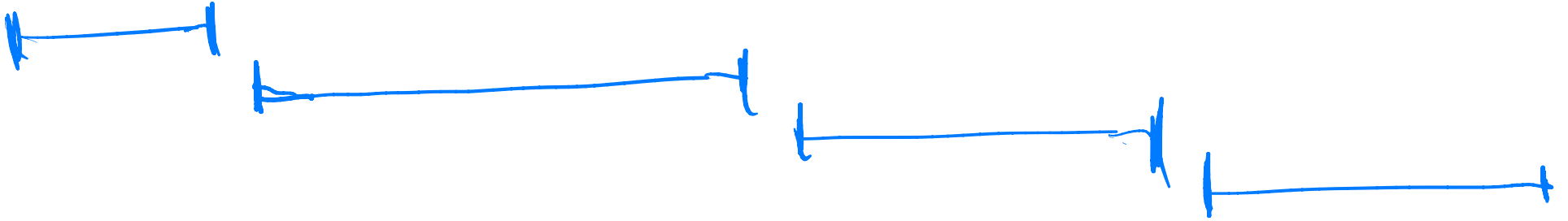
Q_2^{**} : at any moment max 3 activities can overlap

```

RECURSIVE-ACTIVITY-SELECTOR( $s, f, k, n$ )
1   $m = k + 1$ 
2  while  $m \leq n$  and  $s[m] < f[k]$  // find
3     $m = m + 1$ 
4  if  $m \leq n$ 
5    return  $\{a_m\} \cup$  RECURSIVE-ACTIVITY-SELECTOR( $s, f, m, n$ )
6  else return  $\emptyset$ 
  
```

$$Q[s_i: s_n, f_i: f_n] = \underset{\text{value}}{i = \text{greedy choice}} \arg \min(f) + C[s_i: s_n, f_i: f_n, \text{overlapping activities with } i]$$

Criteria: min overlap



Coin Change Denominations = $\{d_1, d_2, \dots, d_n\}$

Example $\{1, 5, 10, 25\}$

T cents,

infinite supply

Task: min # coins $sum = T$

T = 24 Greedy: (0, 10, 1, 1, 1, 1) $opt = 6$

Greedy: choose highest coin that fits in T

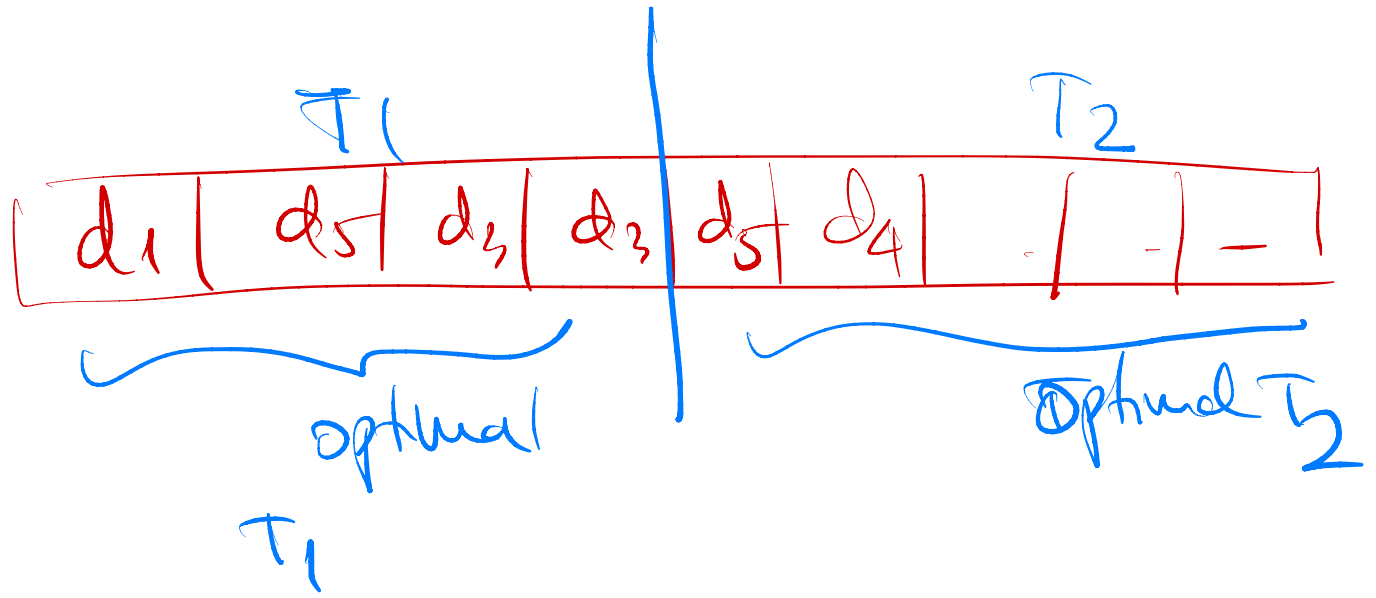
$$C[D, T] = \underset{\text{Greedy } d_i}{1} + C[D, T - d_i]$$

D = $\{1, \del{3}, 4, 5, 7, 10, 25\}$
non-greedy

***: What D (denominations sets) allow greedy solutions?

General case

OPT SOL T



Gas Stations

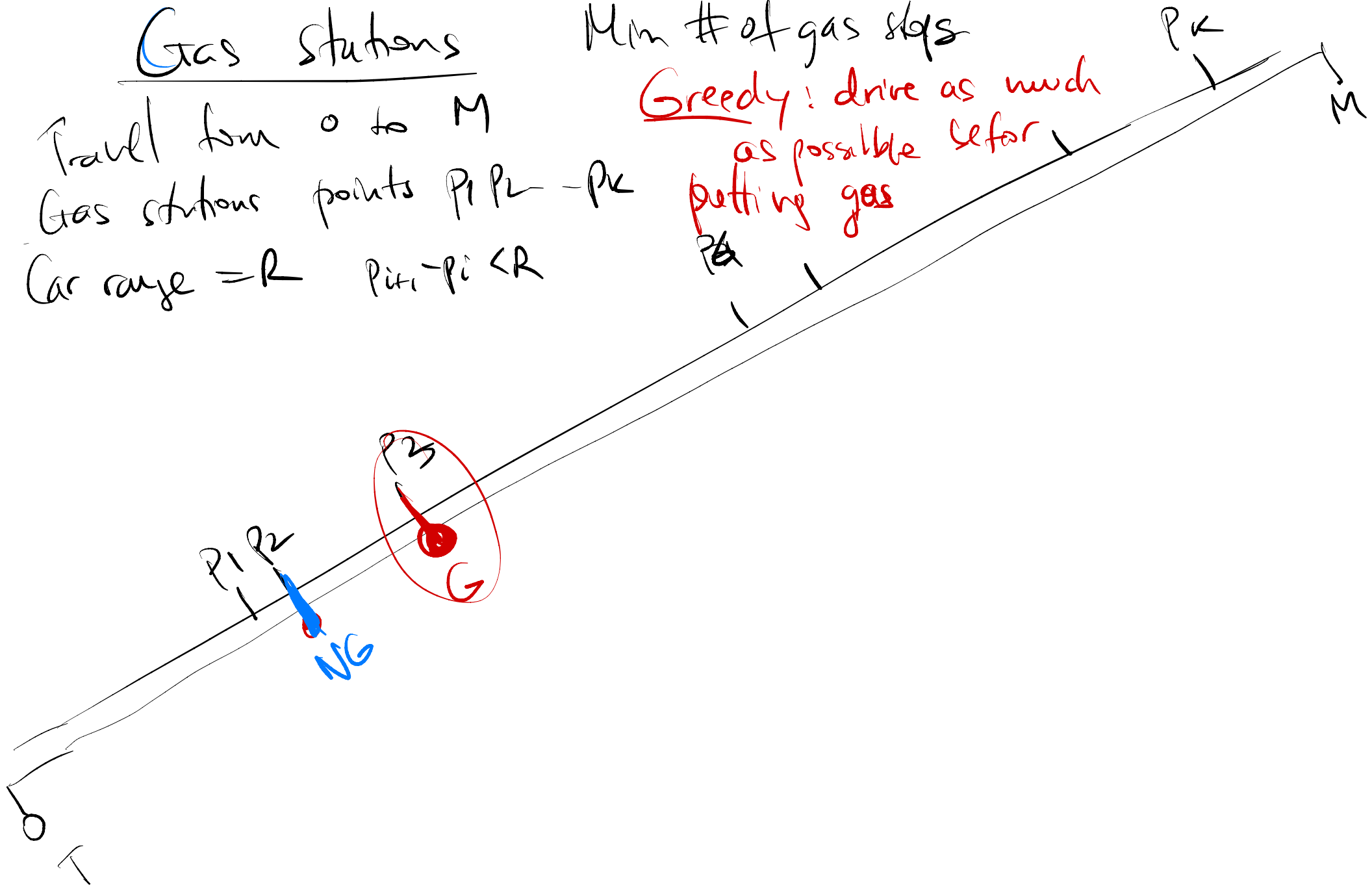
Min # of gas stops

Travel from 0 to M

Gas stations points $P_1 P_2 \dots P_k$

Car range = R $P_{i+1} - P_i < R$

Greedy: drive as much as possible before putting gas



(1h) Any fraction $\in \mathbb{Q} = \sum$ Egyptian fractions

$$\text{egypt. fraction} = \frac{1}{n \in \mathbb{N}}$$

CS task: decompose

$$\text{fraction } \frac{a}{b} = \sum_{i=1}^k \frac{1}{n_i}$$

$$\frac{5}{6} = \frac{1}{2} + \frac{1}{3}$$

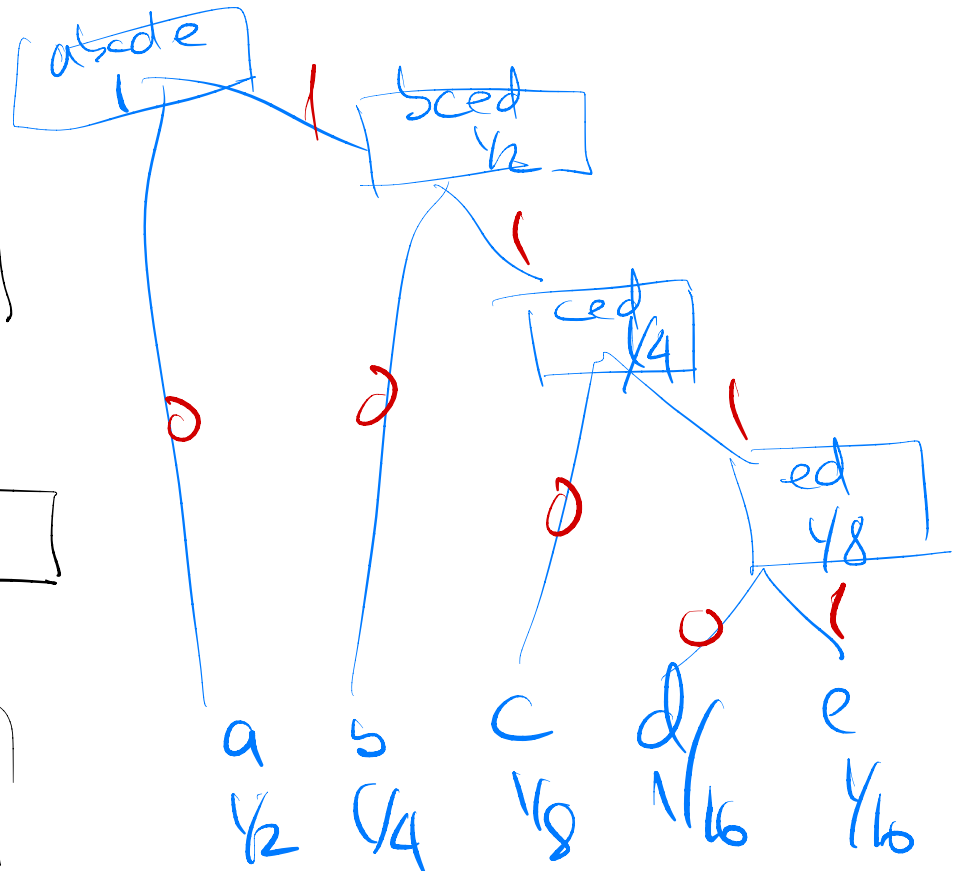
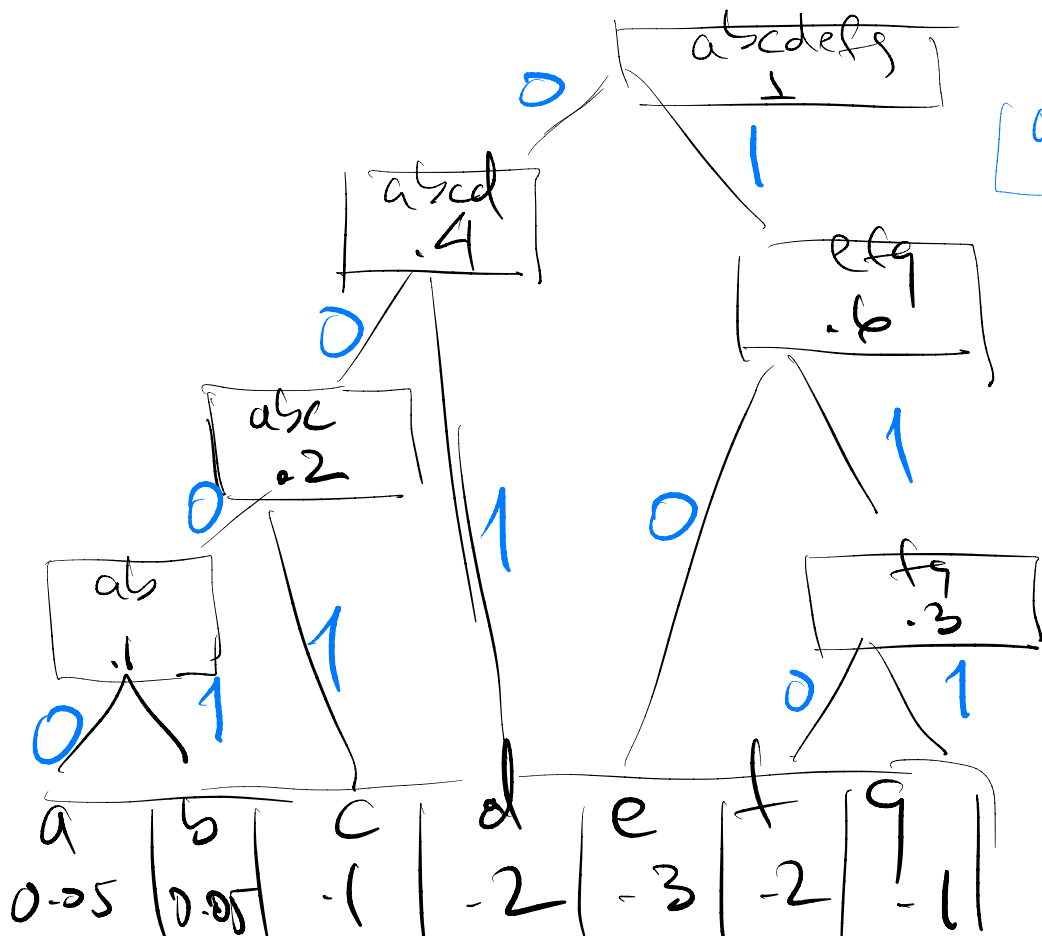
$$\boxed{\frac{7}{15}} = \boxed{\frac{1}{3}} + \boxed{\frac{2}{15}} = \frac{1}{3} + \frac{1}{8} + \frac{1}{120} \quad \checkmark$$

↓
greedy
criteria

subpb = smaller numerator

$$\frac{5}{121} = (\text{best}) \quad \frac{1}{33} + \frac{1}{121} + \frac{1}{363}$$

Greedy: $\frac{5}{121} = \frac{1}{25} + \frac{1}{757} + \frac{1}{763309} + \frac{1}{65} + \frac{1}{55} \dots$



Huffman Codes : Task encode char \rightarrow string of bits
var. length

given char probab

a	b	c	d	e	f	g
.05	.05	.1	.2	.3	.2	.1

ex2

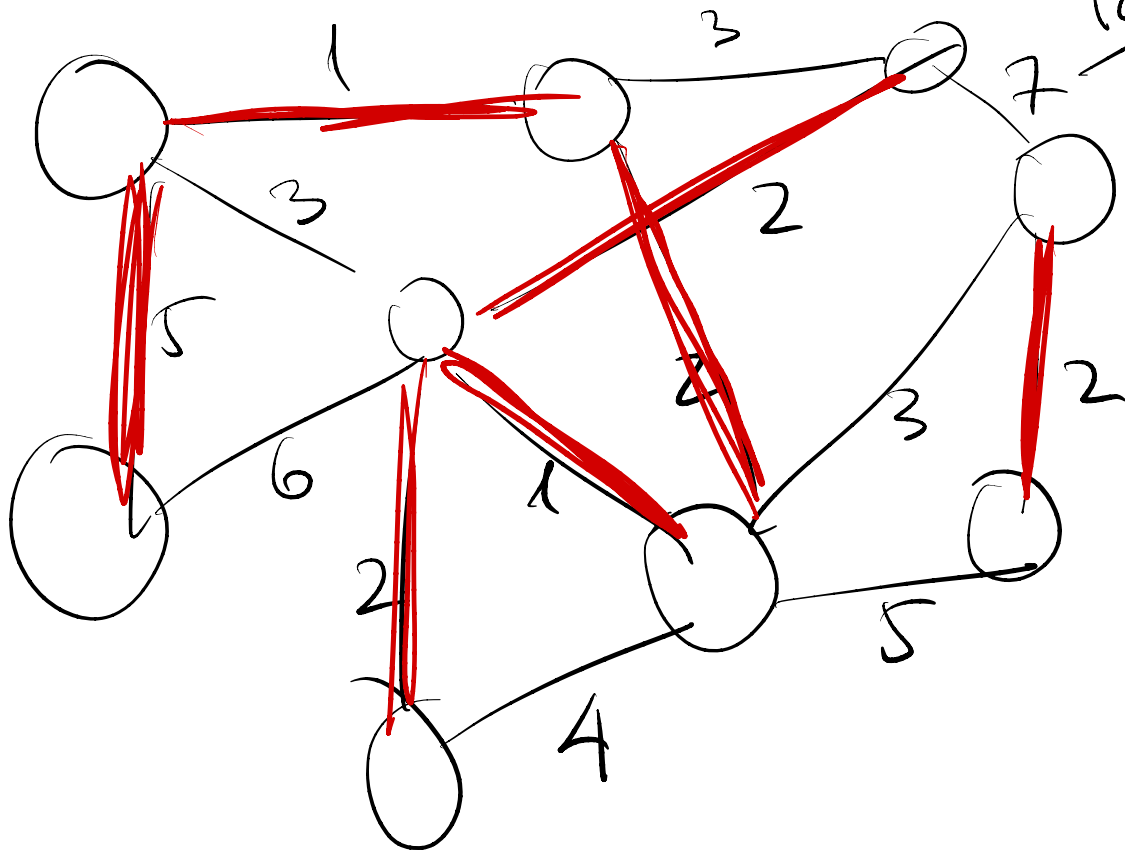
a	b	c	d	e
$\frac{1}{2}$	$\frac{1}{4}$	$\frac{1}{8}$	$\frac{1}{16}$	$\frac{1}{16}$

Obj : minimize $E[\# \text{ of bits/char}]$ in encoding.

$$= \sum_i P(c_i) \cdot |\text{encode length}|_{c_i}$$

constraint : possible to decode "prefix-free code"

MST : Undirected graph $G=(V, E)$ edge weight



Task find subset of edges $\subset E$

- connects all vertices
- no cycles \Rightarrow tree
- total sum min.