

★ Inverse FFT

$$C(\omega) = A(\omega) \cdot B(\omega)$$

$$\therefore C(\omega_{j,2n}) = \sum_{i=0}^{2n-1} c_i(\omega_{j,2n})^i$$

(1)

$$1 \quad \omega_{0,2n} \quad \omega_{0,2n}^2 \quad \dots \quad \omega_{0,2n}^{2n-1}$$

(2)

$$1$$

(3)

$$\vdots$$

(4)

$$1$$

$$1 \quad \omega_{2n-1,2n} \quad \dots \quad \omega_{2n-1,2n}^{2n-1}$$

(Vandermonde matrix)

$$★ V_{jk} = \omega_{j,2n}^k = \omega_{jk,2n}$$

$$V_{jk}^{-1} = -\omega_{jk,2n}^{\frac{1}{2n}} \quad \left. \right\} \text{Claim}$$

$$★ \text{We know } VV^{-1} = I = V^{-1}V$$

$$\text{i.e. } [VV^{-1}]_{ii} = 1$$

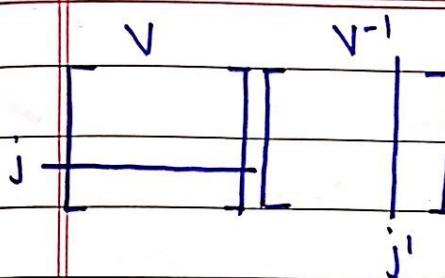
$$[VV^{-1}]_{ij} = 0 \quad (i \neq j)$$

★ Lemma: For $m \geq 1$ and $k \in \mathbb{Z}^+$ which is not a multiple of m , $\sum_{j=0}^{m-1} (\omega_{k,m})^j = 0$

$$\text{Proof: } \sum_{j=0}^{m-1} (\omega_{k,m})^j = \frac{(\omega_{k,m})^m - 1}{\omega_{k,m} - 1}$$

$\because k$ is not a multiple of $m \therefore$ denominator is not zero. Hence $(\omega_{k,m})^m - 1 = 0 \Rightarrow$ Given sum = 0

↓
mth root of unity



grouping bounded
multiplication roundoff

$$[VV^{-1}]_{jj'} = \sum_{k=0}^{2n-1} V_{jk} \cdot V^{-1}_{k,j'} \text{ is bounded}$$

above all the terms V_{jk} and $V^{-1}_{k,j'}$ are bounded

$$= \sum_{k=0}^{2n-1} (w_{jk, 2n} - w_{j', k, 2n})$$

($w_{jk, 2n}, w_{j', k, 2n}$ are bounded)

$$= \frac{1}{2^n} \sum_{k=0}^{2n-1} (w_{k, 2n})^{j-j'} \downarrow$$

is bounded

∴ If $j=j'$ Then $[VV^{-1}]_{jj'} = 1$

$\forall j, k = k$

If $j \neq j'$

$\forall j, k = k$

$$[VV^{-1}]_{jj'} = \frac{1}{2^n} \sum_{k=0}^{2n-1} (w_{(j-j'), 2n})^k$$

* Hence By using the lemma, $\because 2n \geq j-j' \Rightarrow 0$

$$\therefore [VV^{-1}]_{jj'} = 0$$

$\therefore 0 \leq 0 \leq 0$

$$\therefore V^{-1}_{jk} = -\frac{w_{jk, 2n}}{2^n}$$

$\forall j, k = k$

$$y_k = \sum_{j=0}^{2n-1} a_j \cdot w_{kj, 2n}$$

$\exists (y_j)$

$$c_k = \sum_{j=0}^{2n-1} y_j - \frac{w_{kj, 2n}}{2^n} \quad \left. \begin{array}{l} \text{Can be calculated using} \\ \text{FFT (inverted version)} \end{array} \right\}$$

in $O(n \log n)$

$\therefore O(n \log n)$

* Repeated Squaring → Modular exponentiation

$$x^y \bmod n = \begin{cases} (x^{L^{\lfloor y/2 \rfloor}} \bmod n) (x^{L^{\lfloor y/2 \rfloor}} \bmod n) \bmod n & \text{If } y \text{ even} \\ (x^{L^{\lfloor y/2 \rfloor}} \bmod n) (x^{L^{\lfloor y/2 \rfloor}} \bmod n) x \bmod n & \text{If } y \text{ odd} \end{cases}$$

No. of recursive calls: $O(\log y)$

$$T.C = O(\log y \cdot \max(\log x, \log y, \log n))$$

\downarrow
RAM model

* $y = y_0 \text{ LSB}$

$$x^y = x^{y_0 \text{ LSB}}$$

If $\text{LSB} = 0$ then y is even, else y is odd

* int power(int x, int y, int n)

{

int res = 1;

$x^y = n$

while (y) {

if ($y \neq 1$)

res = (res * x) % n;

while ($y >= 1$) {

($x = (x * x) \% n;$) }

return res;

}

* Greedy Algorithm

① JOB SCHEDULING

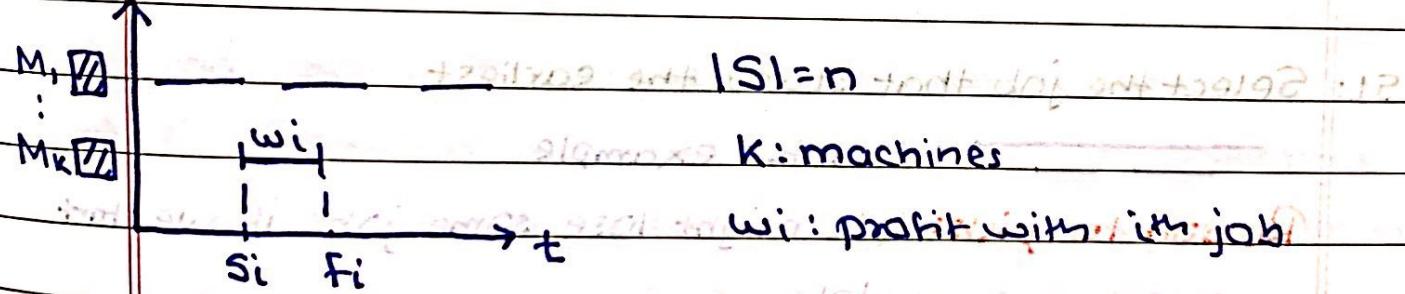
Homogeneous machines: same performance

Heterogeneous machines: different performance

* **No preemption:** Machine is occupied from the start time of the job till the job ends.

* **Mutual compatibility:** No overlapping in jobs

→ Time interval is contiguous.



* Optimal Solution:

- ① Cast the optimization problem as one in which we make a choice and are left with one subproblem to solve.
- ② Check whether it is feasible
- ③ Check the optimality:

By choosing any subset of S you cannot better the solution/profit

* JOB SCHEDULING:

1 Machine, n Jobs (Finish max no. of jobs)

1 2 3 4

s c f

5 6 7 8 9 10 11

$S^1 = \{1, 2, 3, 4\}$: Optimal & feasible (Optimal Solution)

$S^{11} = \{1, 3, 4\}$: Only feasible (Optimal Solution)

$S^{111} = \{1, 2, 3, \dots, 11\}$: Optimal but not feasible

S1: Select the job that starts the earliest

: Counter example

Reason to pick: We might lose some jobs if we start late.

S2: Select the shortest job

: Counter example

Reason: Machine will have more free time

S3: Select the job with fewest conflicts

Counter example: 1st diagram page 11 segment 4

Reason: Need to remove the fewest no. of jobs

S4: Select the job with earliest finish time.

Reason: Machine would be freed earlier.

① Maintain two arrays (sorted by start times)

(One with jobs sorted in finish time order, and

second with the corresponding start times of array 1)

f_1, f_2, \dots, f_n (sorted)

s_1, s_2, \dots, s_n

Greedy Strategy: Earliest finish time

Start at earliest job and take next job if it's compatible.

* Induction on no. of jobs in the output set

→ At every juncture of the algorithm, no comparable jobs get removed and every incompatible job is guaranteed to be removed after placing a job in output array.

Induction Basis:

First job with earliest finish time inserted in the set

and pruning is done correctly.

$f_1, f_2, f_3, f_4, f_5, f_6$

$s_1, s_2, s_3, s_4, s_5, s_6$

Let f_1 be the job with earliest finish time

and let $s_2, s_3 < f_1$. Hence we can't choose job 2 or

job 3 in output set. If $s_4 > f_1$, insert job 4 in output

set. Note in case $s_i < f_1$ for $i \geq 4$ it is anyways going to be less than f_4 .

Hence pruned correctly.

Induction Hypothesis:

$$S' = \{j_1, j_2, \dots, j_i, \dots\}$$

upto i^{th} job we have correctly chosen (followed greedy strategy)

Inductive Step

Next job would be chosen with earliest finish time

- * The solution output by greedy algorithm is feasible i.e. every two jobs in S' are compatible.

Let two jobs of S' not be compatible

say j_p and j_r

WLG let $p < r$ i.e. $f_p > s_r$

Consider q such that $p \leq q < r$

such that: $f_q \geq f_p > s_r$

But when we insert j_q in set S'
we must remove j_r . Hence contradiction

* STAYS AHEAD ARGUMENT:

\exists a Optimal Solution $\#$ Naive: $\Omega(2^n) : 0$

Greedy output: S'

Let n' be the number of jobs in S' . For every $1 \leq i \leq n'$ i^{th} job chosen into S' stays ahead of i^{th} job in the optimal solution O when jobs in O are sorted according to their finish times.

Induction on no. of jobs selected in S' :

Induction Base:

$f_1 \leq f_{1'}$ (# Greedy Strategy)

Induction Hypo:

$f_1 \leq f_{1'}, f_2 \leq f_{2'}, \dots, f_k \leq f_{k'}$

Inductive Step:

TPT: $f_{k+1} \leq f_{(k+1)'}$

We can choose $(k+1)^{\text{th}}$ job

such that $f_{k+1} \leq f_{(k+1)'}$

$f_k \quad f_{k'} \quad f_{(k+1)'}$



Hence by Induction S' stays ahead of any optimal solution O .

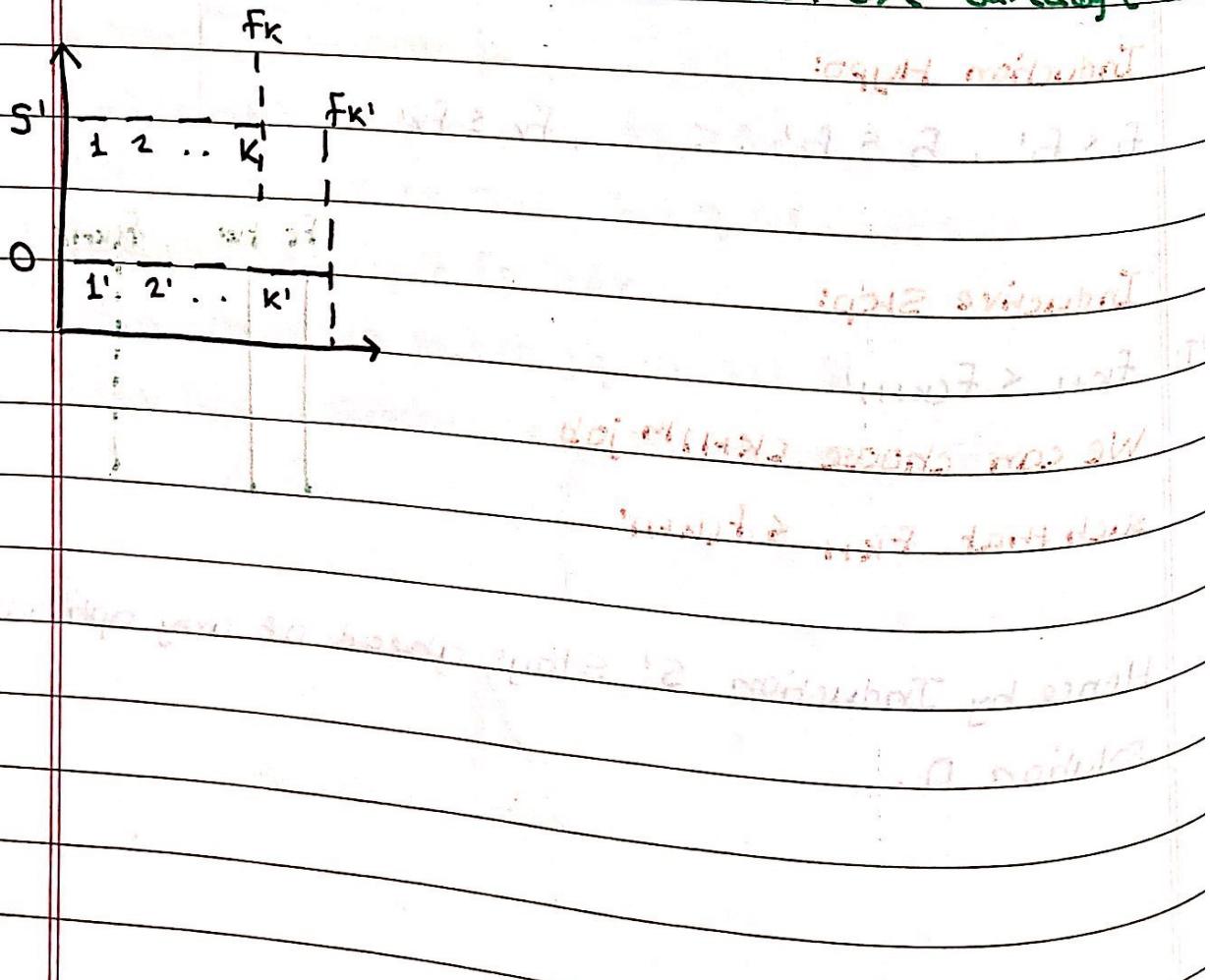
- * Given that the solution output by greedy algo S' 'stays ahead' of any optimal solution O , then S' must be an optimal solution.

If $|S'| < |O|$ consider jobs with ids $|S'| + 1, \dots, |O|$

in the sorted order of jobs in O , let it be O'

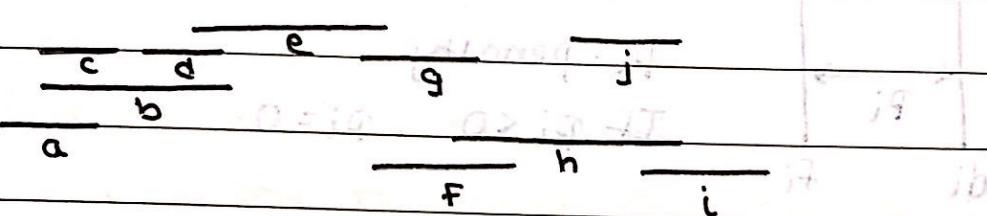
Every job in O' is compatible with every job in S' & hence available for greedy algo after it had chosen set S' of jobs. \rightarrow no conflict

- * Note greedy algo stops only when there are no compatible jobs to the ones which are already chosen



* Find Minimum no. of machines required to schedule all the tasks.

→ Duration of n tasks given



* Pool of machines : S

Assign a machine at the start of the task

If no machine among the pool is available, push a new machine to the pool.

As soon as a task is completed the machine performing that task is free.

* Hence according to input instance any algo should use atleast the maximum no. of conflicting machines.

MEETING THE LOWER BOUND:

Input instance says any algo should use this much amount of minimum resources and our algo matches it.

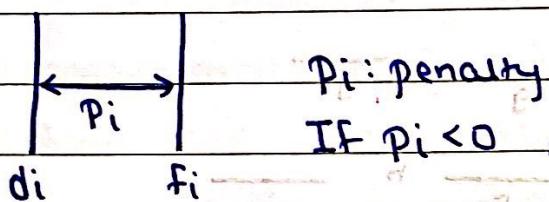
→ Algo: Push all start times & finish times in a vector and sort according to their values & maintain counter for no. of overlaps.

$O(n \log n)$

* Deadlines

Minimise the maximum penalty

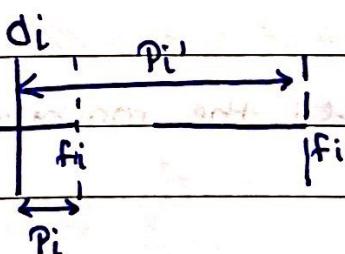
Each task given length & deadline to submit



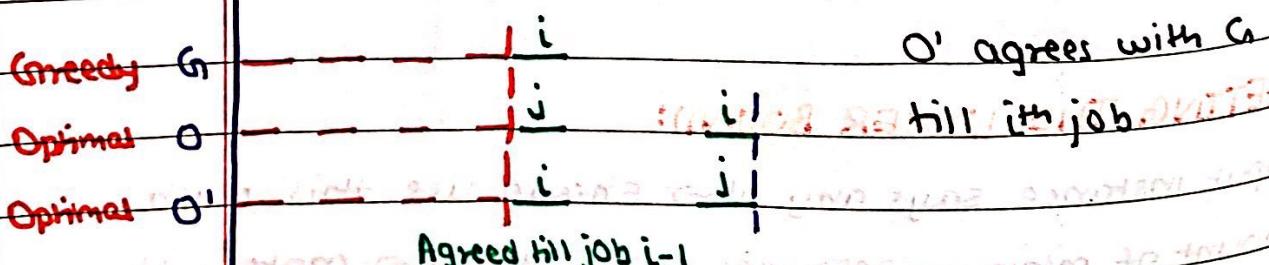
* GREEDY STRATEGY:

Earliest deadline first will give minimum penalty

Reason: More free time to the machine

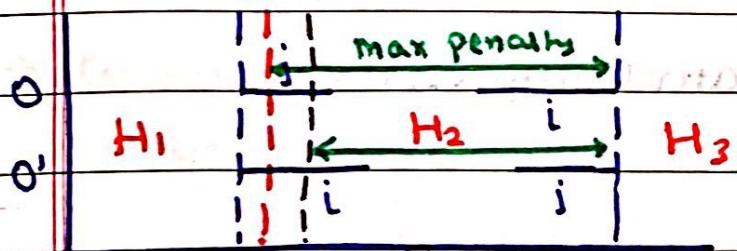


* EXCHANGE ARGUMENT:



$d_j > d_i$ (# Greedy Strategy)

* Optimal solution exists can be found in $O(n!)$

$d_i d_j$ 

$$f_{i \dots d, D} = ?$$

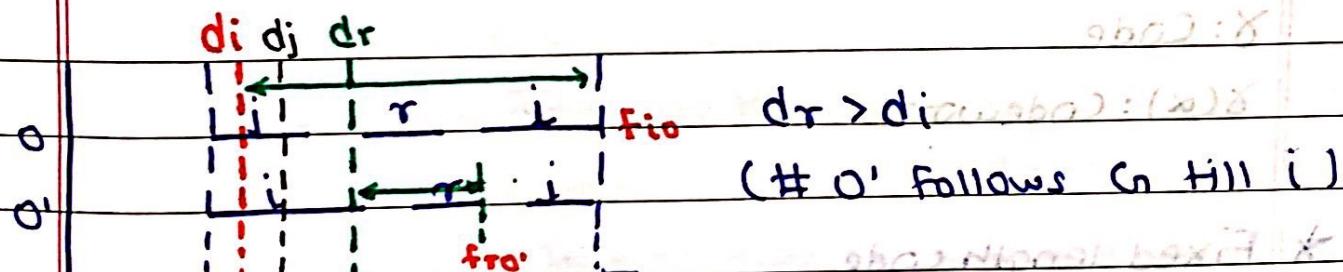
$\therefore \text{max penalty}(0') \leq \text{max penalty}(0)$

* If interval $r \in H_1$

then we have no change in its penalty

Or so for $r \in H_3$

* Let $r \in H_2$



and $f_{i0} > f_{i0'}$

$\therefore \text{max penalty}(0') \leq \text{max penalty}(0)$

* Hence we keep on making optimal solution more coherent with the greedy solution hence our greedy solution is optimal.

* Huffman Codes and Data Compression.

$$S = \{a, b, \dots, z\}$$

(Text) T

... → Symbols from S

* Also freq of each symbol is given as f_a, f_b, f_c, ..., f_z

a b c ... z

f_a f_b f_c ... f_z

γ : Code

$\gamma(x)$: Codeword

* Fixed length code

> 2⁶ < 2⁷ symbols

7 bits for each symbol

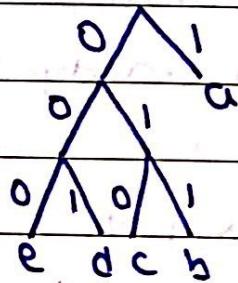
* Avg bits for symbol = $\sum_{x \in S} f_x |\gamma(x)|$

* Prefix code

No code is prefix of any other code.

a: 11 b: 01 c: 001 d: 10 e: 000

- * Given a binary tree + leaves associated with symbols
 → prefix code



a: 1 $f_a = 0.32$
 b: 011 $f_b = 0.25$
 c: 010 $f_c = 0.20$
 d: 001 $f_d = 0.1875$
 e: 000 $f_e = 0.05$

→ Proof by contradiction

Let $\gamma(z)$ be prefix of $\gamma(y)$ where $z \neq y$

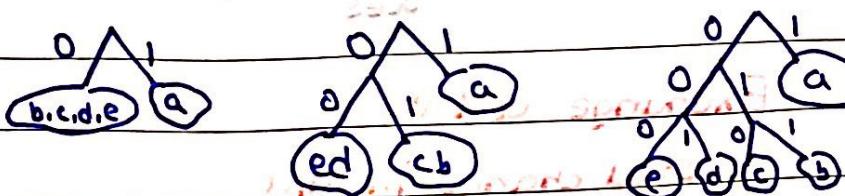
∴ $\gamma(z)$ is prefix of $\gamma(y)$

⇒ z is some node in the path from root to the leaf node y

But z must be a leaf node

- * Prefix code given can construct the tree

a: 1 b: 011 c: 010 d: 001 e: 000



- * Binary Tree corresponding to optimal prefix code is full.

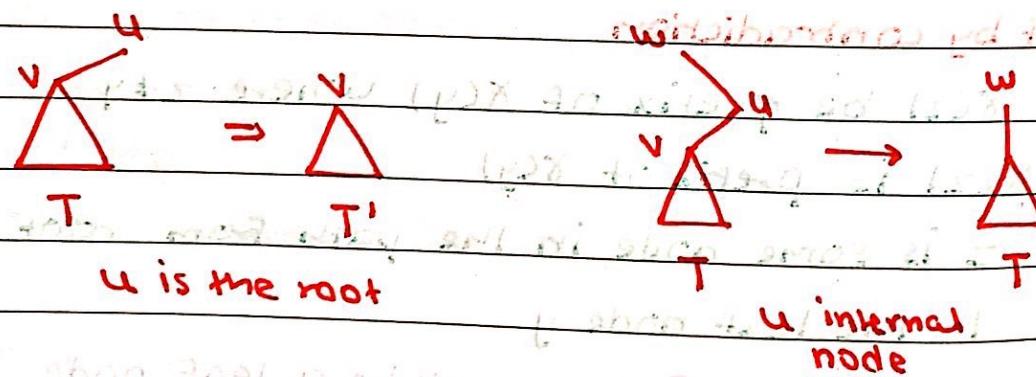
Proof by Exchange Argument

Let T be the binary tree corresponding to optimal prefix code and \exists a node u with exactly one child v .

Now convert $T \rightarrow T'$ by replacing $u \& v$

$\therefore T'$ has smaller avg no. of bits.

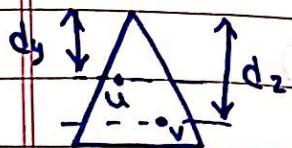
contradicts optimality of T



- * δ^* optimal corresponds to T^* & u, v be leaves
 $\exists \text{depth}(u) < \text{depth}(v)$ If u comes, $y \in S$ &
 v comes, to $z \in S$ then $f_y \geq f_z$

Proof by Exchange argument

$$\text{Avg Bits per symbol (ABS)} = \sum_{x \in S} f_x \text{depth}(x)$$



Exchange $u \& v$

Overall change in ABS

$$\text{i.e. } \Delta \text{ABS} = (\text{depth}(v) - \text{depth}(u))(f_y - f_z)$$

If $f_z > f_y \therefore \Delta \text{ABS} -ve \# \text{ to optimality of } T^*$

* There is an optimal prefix code comes to T^* where two lowest freq letters are assigned to leaves that are siblings in T^*

→ The choice of assignment among leaves at same depth doesn't affect the avg no. of bits per symbol.

* Algo for Optimal prefix code:

Let y^* & z^* be having two lowest freq in S^*

parent w^* : meta letter with freq

$$f_w^* = f_{y^*} + f_{z^*}$$

Replace y^* & z^* with this w^*

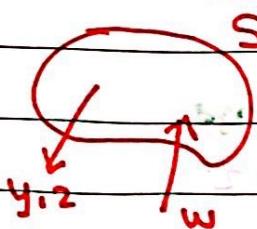
$$\text{ABS}(T') = C + f_{w^*} d_{T'}(w^*)$$

$$\text{ABS}(T) = C + f_{y^*} d_T(y^*) + f_{z^*} d_T(z^*)$$

$$= C + f_{w^*} (d_{T'}(w^*) + 1)$$

$$\therefore \text{ABS}(T) - \text{ABS}(T') = f_{y^*} + f_{z^*} = f_{w^*}$$

* Huffman's prefix code works on T^* & S^*



$$S' = S - \{y, z\} \cup \{w\}$$

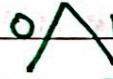
* Binary tree # at every step two symbols removed & hence every node has two children (internal node)

* Optimality

→ Induction on no. of symbols

Induction Basis

For 2 symbols



Optimal # only one bit per symbol

Induction Hypo.

$|S| = k-1$ (Optimal)

T.P.T. $|S| = k$ also! Optimal

Induction Step

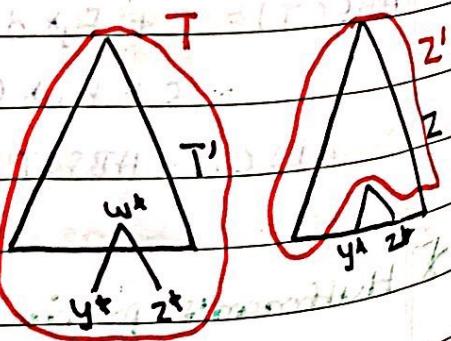
Let S' be set of symbols $\{y^*, z^*\}$

where $\{y^*, z^*\}$ replaced by w^* which is optimal I.H.

Let the tree be T'

attach leaves y^* & z^* to w^*

$$ABL(T') = ABL(T) - fw$$



Now let T not be optimal

$$\therefore \exists Z \ni ABS(Z) < ABS(T)$$

there is a tree $Z \ni \{y^*, z^*\}$ are siblings

If we delete y^*, z^* from Z to get Z'

$$\therefore ABS(Z') = ABS(Z) - fw$$

$$\therefore ABS(Z) < ABS(T)$$

$$\Rightarrow ABS(Z') < ABS(T') \# \text{to optimality of } T'$$

* Heap with n symbols

extract two with min freq

& insert their sum

Build $O(n)$

Extract $O(n \log n)$

Insert $O(n \log n)$

$0+0 | 1 | 1 | 0$

$0+1 | 0 | 0 | 1 | 0$

$0+0 | 1 | 1 | 0$

* T.C. $O(n \log n)$

$0+1 | 0 | 0 | 1 | 0$

* Greedy Choice Property:

Many choices choose the one which looks best at the moment.

* Optimal Substructure Property

Optimal Solution to problem P within it contains

optimal substructure (solutions) to subproblems of P.

Labour cost brief : $\text{labour cost} = \text{labour cost} + \text{labour cost}$

$\rightarrow \text{labour cost}$

and if we add one more labour cost then total cost will increase.

we consider a S algorithm

$S \in M \rightarrow \text{labour cost} = \text{labour cost} + \text{labour cost}$

$\rightarrow M$

$\text{labour cost} = \text{labour cost} + \text{labour cost}$

* Matrix Exponentiation (Repeated Squaring)

$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} \rightarrow f_0 \rightarrow f_1$$

A

$$\begin{bmatrix} 0 & 1 \\ 1 & 1 \end{bmatrix}^n \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} f_{n-1} \\ f_n \end{bmatrix}$$

Proof By
Induction

$$A^n = \begin{cases} A^{\lfloor \frac{n}{2} \rfloor} \cdot A^{\lfloor \frac{n}{2} \rfloor} & \text{if } n \text{ is even} \\ A \cdot A^{\lfloor \frac{n-1}{2} \rfloor} \cdot A^{\lfloor \frac{n-1}{2} \rfloor} & \text{if } n \text{ is odd} \end{cases}$$

$$T(n) = T(n/2) + O(1) \quad \text{if } n > 2$$

$$T(1) = O(1) \quad \text{Initial condition}$$

$$\therefore T(n) = O(\log n) : \text{Word ram model}$$

$$f_n \approx \phi^{n-1}$$

$$\therefore n \log \phi \approx O(n) : \text{no. of bits of } n^{\text{th}} \text{ Fibonacci no.}$$

$$\therefore T(n) = T(n/2) + M(n) \quad n > 2$$

↑
Multiply 2 n bit nos
O(1)

$$M(n) = O(n^{1.59})$$

$$\therefore T(n) = O(n^{1.59})$$