

2 February 2025

Dynamic Programming II

Edit Distance

Plan

- * Review of Dynamic Programming
- * Announcements
- * Sequence Alignment Problem
(aka Edit Distance)

Dynamic Programming

* Search for optimal solution by exploiting sub-structure.

Example. Max Wt Independent Set on Paths



$$WIS(P_n) = \max \left\{ \begin{array}{l} WIS(P_{n-1}), \\ w_n + WIS(P_{n-2}) \end{array} \right\}$$

DP Recurrence \Rightarrow Recursive Algorithm

Key Ingredient: Record Answers as you go!
↳ Avoid Redoing work.

$W = [-1, -1, \dots, -1]$ // Dynamic Programming Table

Memoized WIS (P_k):

if $k=0$, return 0

if $k=1$, return W_1 .

if $W[k-1] = -1$:

| $W[k-1] \leftarrow \text{Memoized WIS } (P_{k-1})$

if $W[k-2] = -1$:

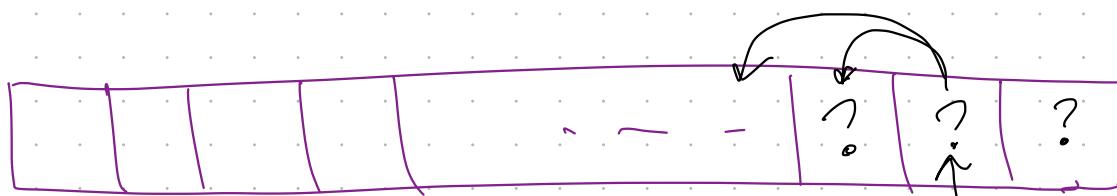
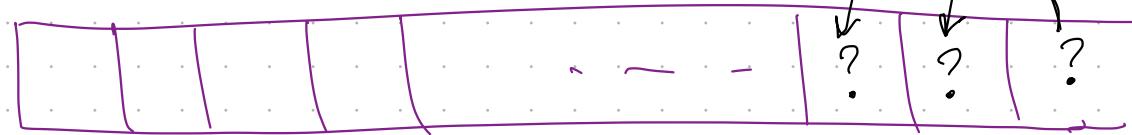
| $W[k-2] \leftarrow \text{Memoized WIS } (P_{k-2})$

return $\max \{ W[k-1], W_n + W[k-2] \}$

$W =$



Memoized WIS(P_n)



Memoized WIS(P_{n-1})

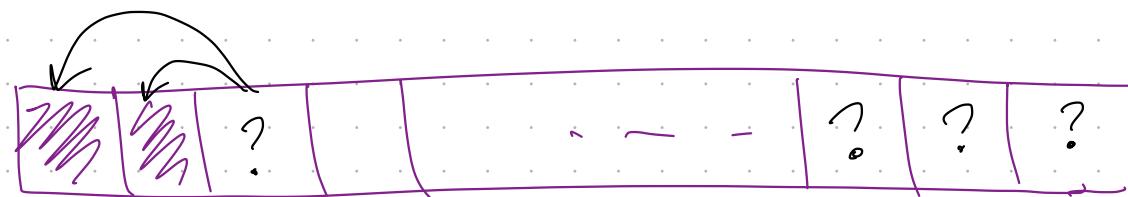
$W =$



Memoized WIS(P_n)



Memoized WIS(P_{n-1})



Base Cases P₀ P₁

Once Base Cases
Return

DP Table fills
up in linear cascade.

$W =$



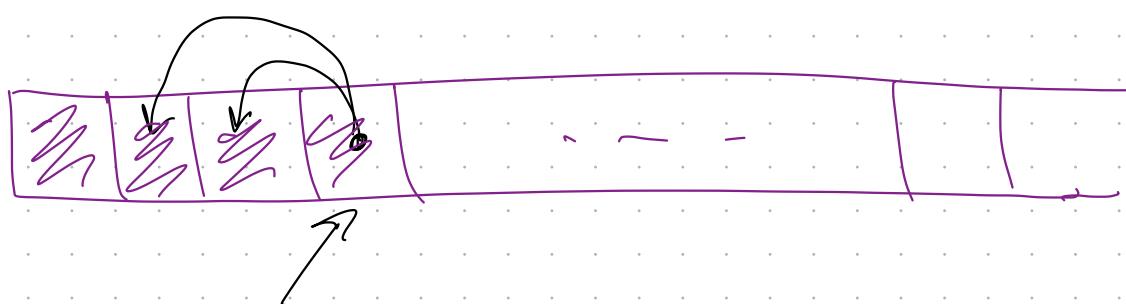
Memoized WIS(p_n)



Base Cases $P_0 \ P_1$

Once Base Cases
Return

DP Table fills
up in linear cascade.



$$W[k] = \max \{ W[k-1], w_n + W[k-2] \}$$

Iterative Solution

- * Recursive Formulation conceptually nice.
- * Dynamic Programs always have an equivalent iterative formulation

→ fills in the DP Table directly

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Iterative WIS (P_n) .

Let $W = [-1, -1, \dots, -1]$

$W[0] \leftarrow 0$, $W[1] = w_1$.

for $k = 2, \dots, n$:

$W[k] \leftarrow \max \{ W[k-1], w_k + W[k-2] \}$

Return $W[n]$

From here on,
we write the iterative
version

Announcements

- * HW 0 Grades Released After lecture
- * HW 1
- * Exam conflict Survey
 - ↳ Posted to Ed.
 - ↳ Due Thurs.
 - ↳ Only for University-approved conflicts

Proofs of Correctness.

- * What does it mean to prove algorithm A is "correct" for problem Π ?

For all instances x of problem Π ,
 $A(x)$ returns the right answer.

Proofs of Correctness.

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For all instances x of problem Π ,
 $A(x)$ returns the right answer.

- * Be precise & Break Correctness in pieces.

↳ e.g.  connected undirected graphs G , w/ distinct edge weights
weighted Kruskal's returns MST of G .

Show. Kruskal's returns

- a tree T
- T is spanning (i.e. connected)
- T is minimum wt

Dynamic Programming Proofs of Correctness

* Need to prove the Recurrence is correct.

Ex. For all $n \in \mathbb{N}$ and any non-negative weights on the path graph P_n ,

$$WIS(P_n) = \max \left\{ P_{n-1}, w_n + P_{n-2} \right\}$$

OR

Base Cases

Sequence Alignment.

Genomics

* Genome Sequences $S \in \{A, C, G, T\}^*$

$S_{\text{cat}} = A \textcolor{red}{C} \textcolor{blue}{C} \textcolor{purple}{G} \textcolor{red}{A} \textcolor{teal}{T} \textcolor{blue}{C} \textcolor{purple}{G} \textcolor{red}{A} \textcolor{teal}{T} \dots$

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How related are two species?

→ How "similar" are their genomes?

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BLAST: Basic Local Alignment Search Tool > 110 K
citations

Edit Distance

How easily can we edit
string s into string t ?

Edit Distance

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Allowed Edits

- * Insertion / Deletion
- * Changes

EDIT

MINIMUM

SNOW

DIST

MAXIMUM

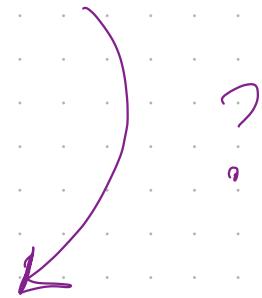
NO

Edit Distance

How easily can we edit string s into string t ?

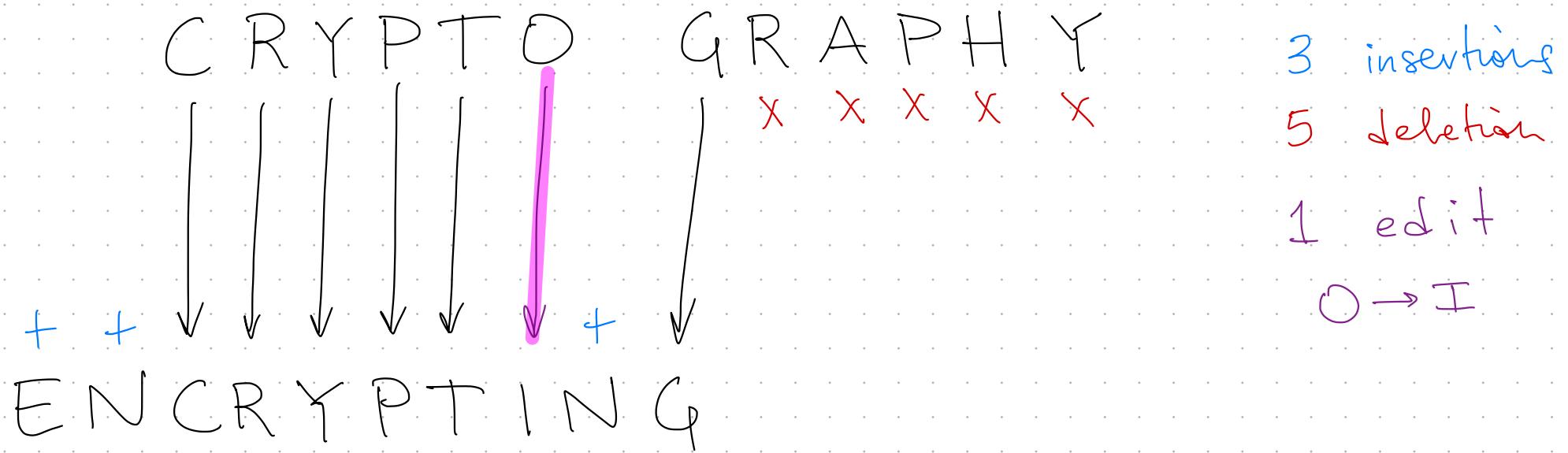
CRYPTOGRAPHY

ENCRYPTING



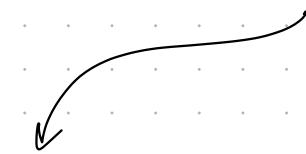
Edit Distance

How easily can we edit string s into string t ?



Given two strings S and T

Compute minimum cost edits from $S \rightarrow T$



insertion / deletion : γ

"changing" a to b : Δ_{ab}

S

AGGCTAAC ---

T

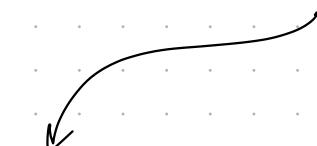
GAGCTAACCC ---



Changing a to a
costs $\Delta_{aa} = 0$.

Given two strings S and T

Compute minimum cost edits from $S \rightarrow T$



insertion / deletion : γ

changing a to b : Δ_{ab}

S

AGGCTAATC ---



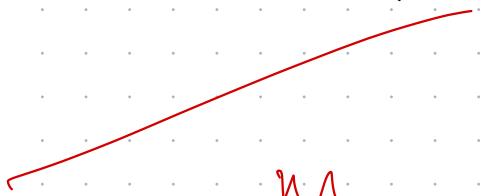
n

T

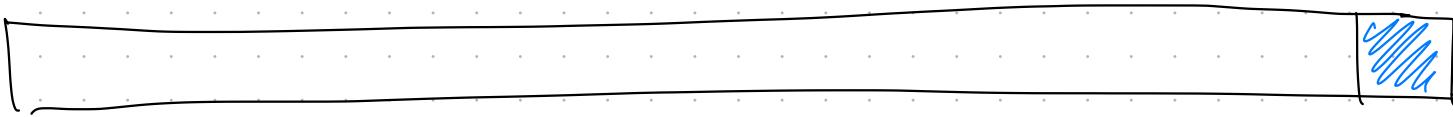
GAGCTAAGCC ---



m



S



S_n

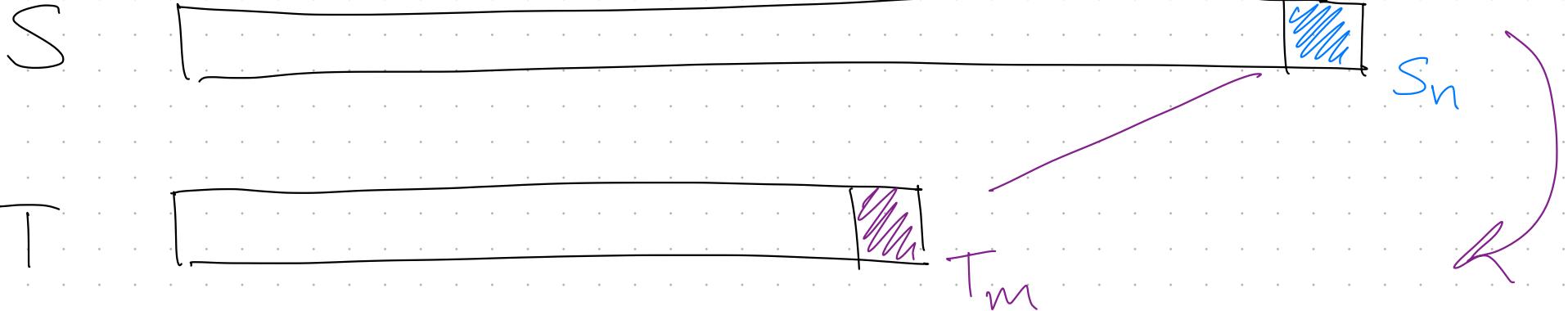
T



T_m

Fact. Fix an optimal set of edits from S to T.

One of the following is true:



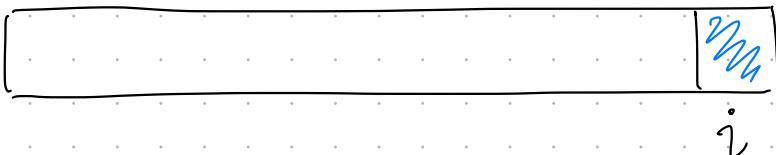
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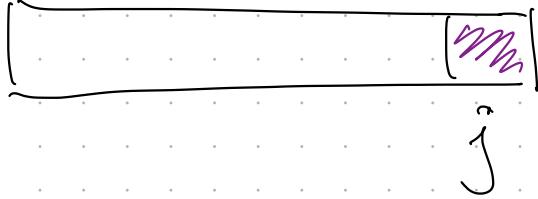
- * S_n is deleted at cost \times
- * T_m is inserted at cost γ
- * S_n "changes" to T_m at cost $\Delta_{S_n T_m}$

The Edit Distance Recurrence

$S[1:i]$



$T[1:j]$



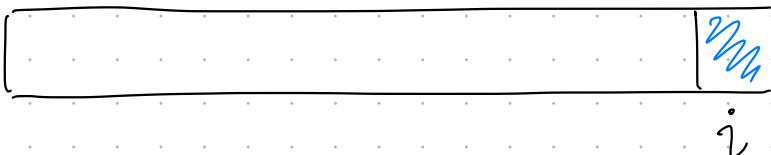
$ED(i, j)$



Edit distance
between prefixes
 $S[1:i]$ and $T[1:j]$

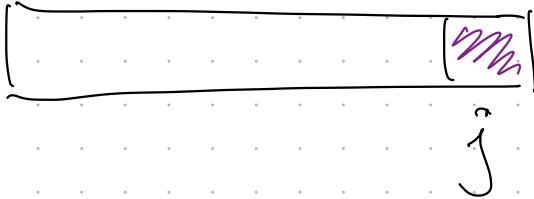
The Edit Distance Recurrence

$S[1:i]$



i

$T[1:j]$



j

$ED(i, j)$

Possibilities

- * S_i deleted $\longrightarrow ED(i-1, j) + \gamma$
- * T_j inserted $\longrightarrow ED(i, j-1) + \gamma$
- * S_i changes to $T_j \longrightarrow ED(i-1, j-1) + \Delta_{S_i T_j}$

Theorem. The Edit Distance between $S[1:i]$ and $T[1:j]$ is given as

$$ED(i, j) = \min \left\{ \begin{array}{l} ED(i-1, j-1) + \Delta_{S_i T_j}, \\ ED(i-1, j) + \gamma, \\ ED(i, j-1) + \gamma \end{array} \right\}$$

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Base Cases?

$$ED(i,0) = i \cdot \gamma \rightarrow$$

$$ED(0,j) = j \cdot \gamma$$



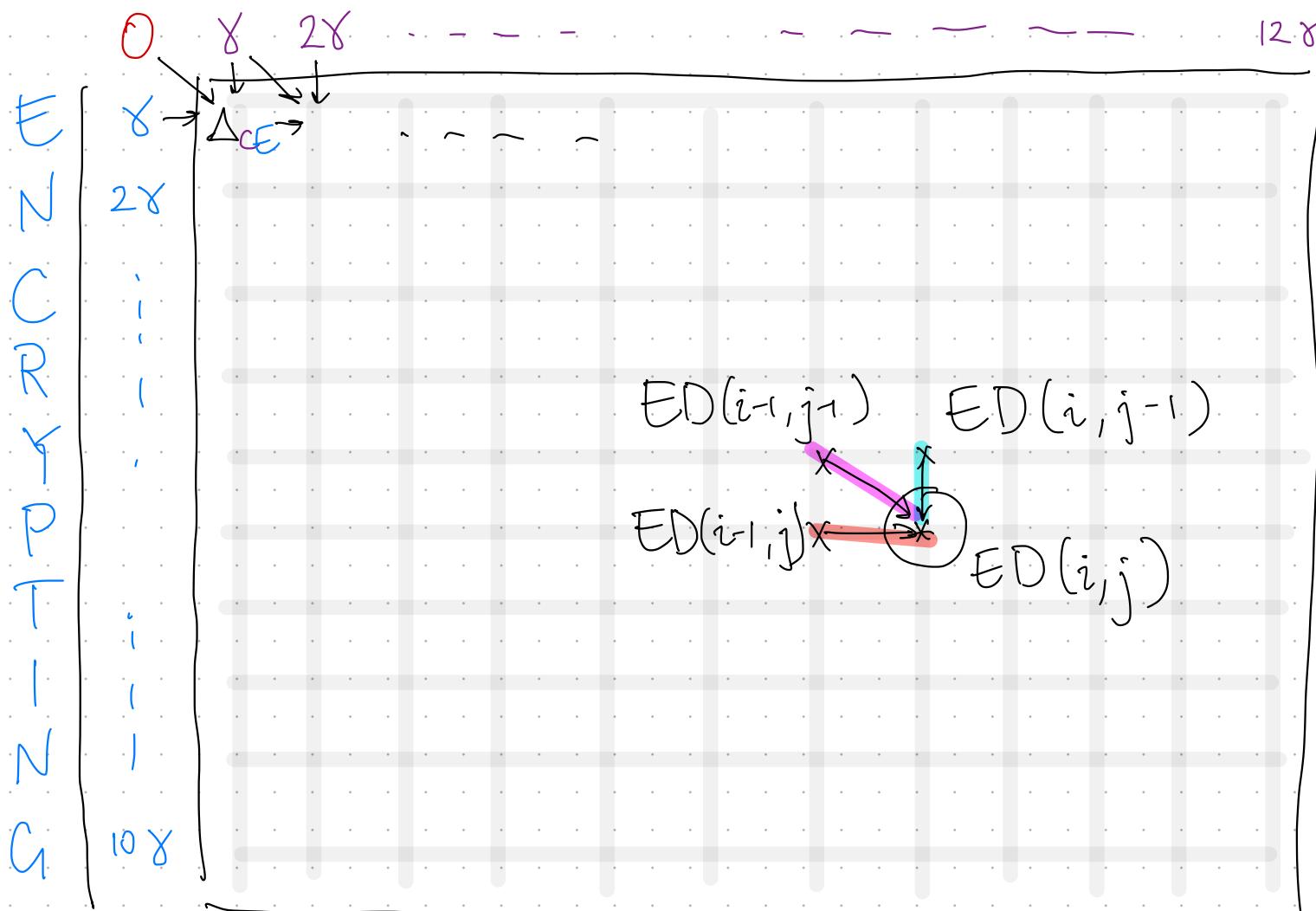
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$ED(i, j)$

2D dynamic programming
table

C R Y P T O G R A P H Y



Each entry requires 3 probes into prior entries.

Edit Distance Algorithm

$$ED(0,0) = 0$$

$$ED(i,0) = i \cdot \gamma$$

$$ED(0,j) = j \cdot \gamma$$

For $i = 1 \rightarrow n$

 For $j = 1 \rightarrow m$

$$ED(i,j) = \min \left\{ \begin{array}{l} ED(i-1, j-1) + \Delta_{S_i T_j}, \\ ED(i-1, j), \\ ED(i, j-1) + \gamma \end{array} \right\} + \gamma$$

return $ED(n,m)$

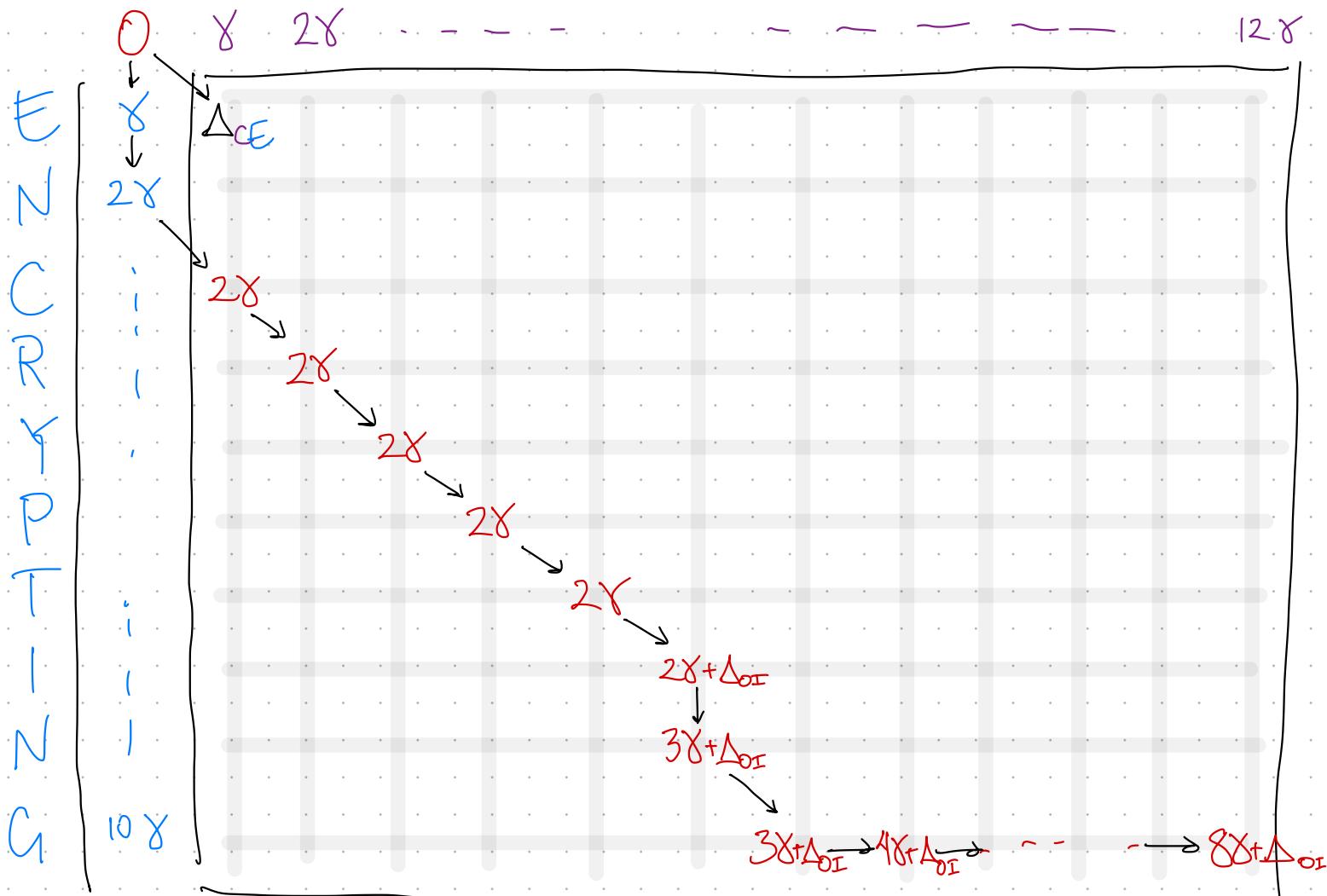
insertions

Edit

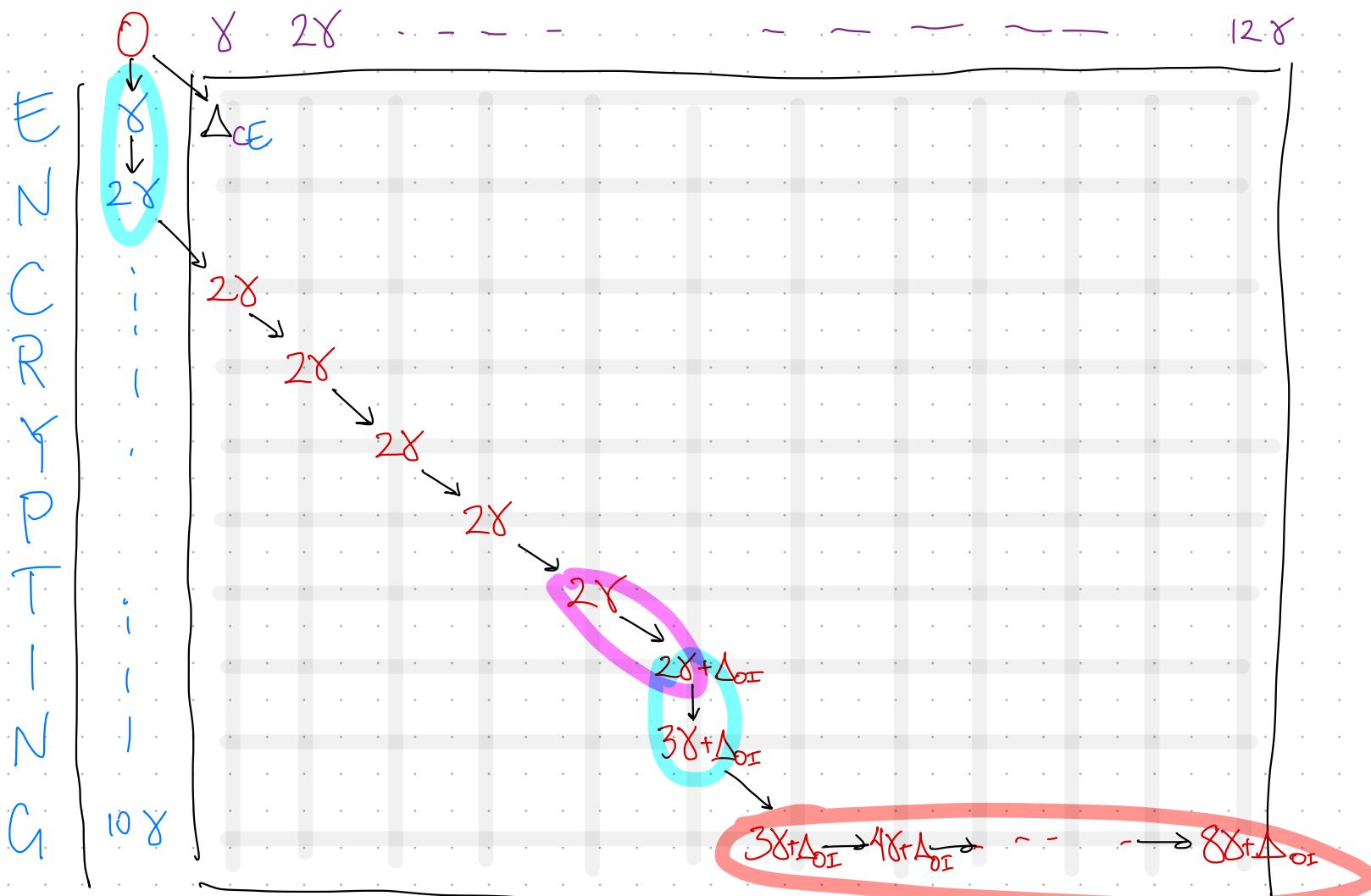
++ CRYPTOGRAPHY
|||
ENCRYPTION - - - - -

deletions

CRYPTOGRAPHY



C R Y P T O G R A P H Y



Edit Distance Algorithm

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Running Time

- $1 + n + m$ initialization

- n outer iterations

m inner iterations

$O(1)$ per iteration

For $i = 1 \rightarrow n$

For $j = 1 \rightarrow m$

$$ED(i,j) = \min \left\{ \begin{array}{l} ED(i-1, j-1), ED(i-1, j), ED(i, j-1) \\ + \Delta_{S_i T_j} \quad \quad \quad + \gamma \quad \quad + \gamma \end{array} \right\}$$

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Space

DP Table has $(n+1) \times (m+1)$ cells

$O(mn)$

Edit Distance \equiv Sequence Alignment

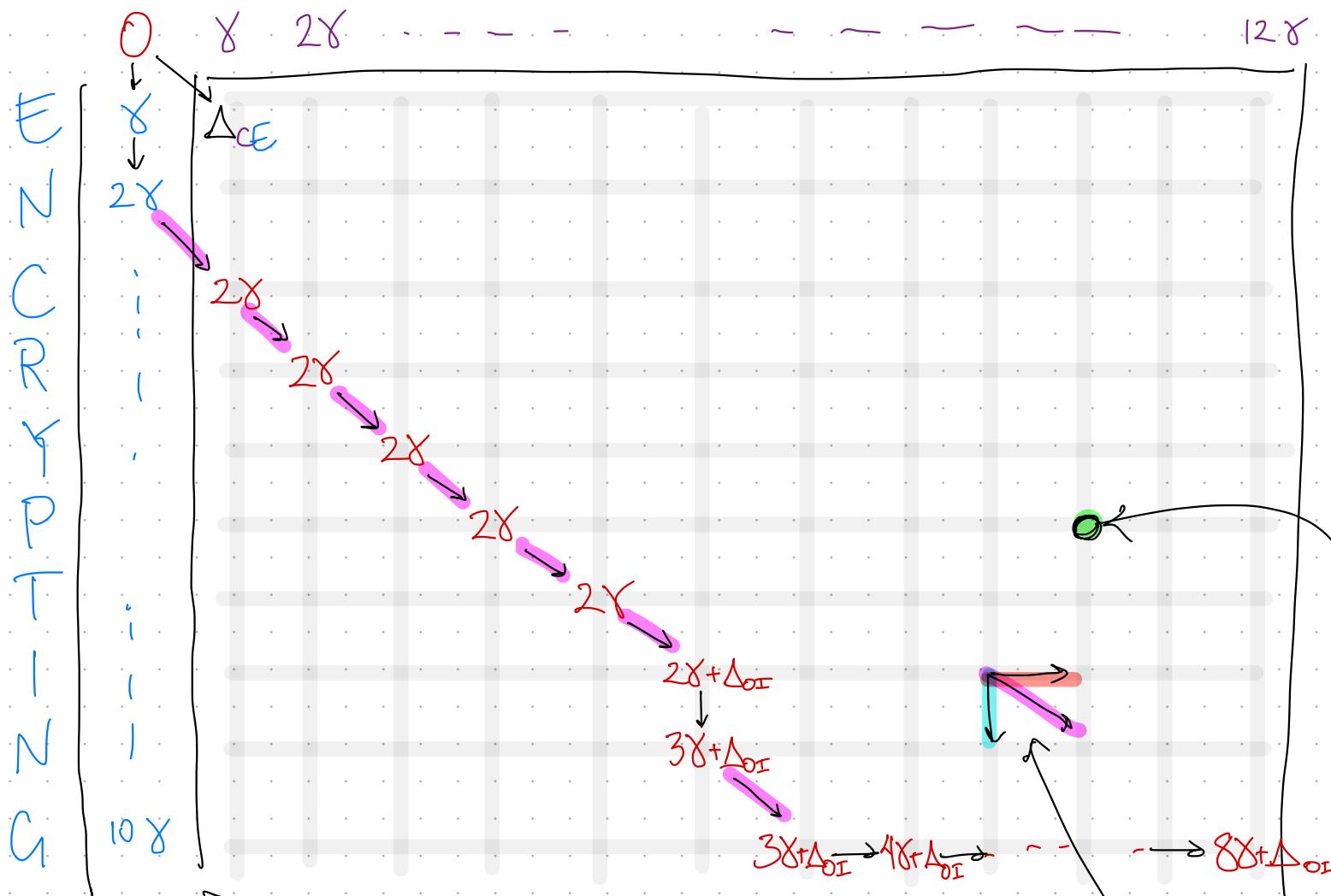
CRYPTOGRAPHY
||| || |
ENCRYPTING

An alignment is a non-crossing matching -

ED Algorithm computed the distance

Can we compute the alignment?

C R Y P T O G R A P H Y



Idea.

Treat
DP Table
as a
graph

entries

=
Vertices

Min ED alignment = shortest path

diagonal edges = alignment.

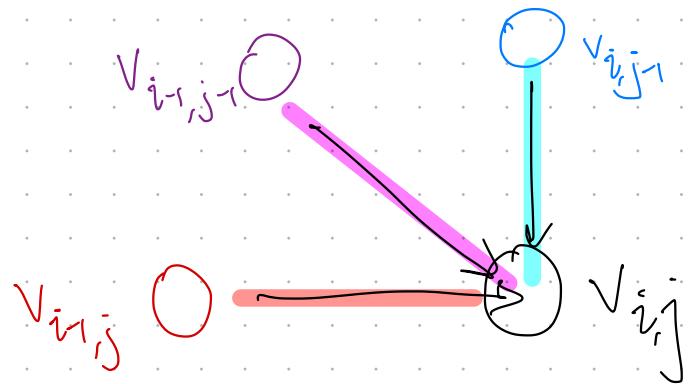
transitions
= edges

Theorem. Shortest path from $v_{00} \rightarrow v_{ij}$
equals $ED(i,j)$.

Pf. By induction on $i+j$.

Base Case. Shortest path from $v_{00} \rightarrow v_{00} = 0$
 $= ED(0,0)$

Inductive Step. Suppose shortest path to $v_{kl} = ED(k,l)$
for all $k+l < i+j$.



Consider SP to v_{ij} .

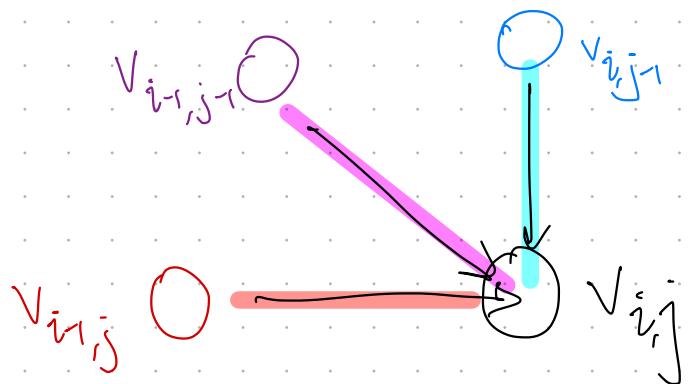
↳ only 3 in-edges.

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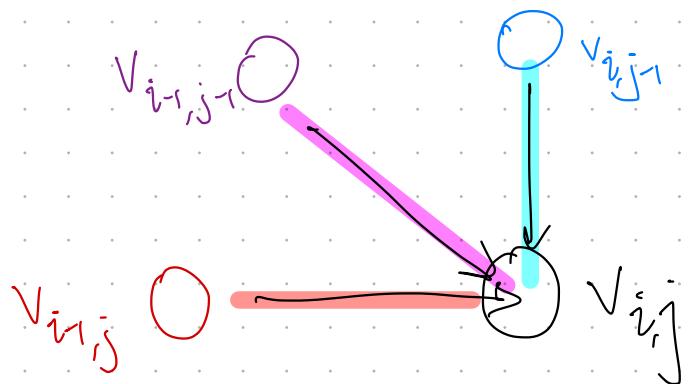
$$SP(v_{00} \rightarrow v_{ij}) = \min \left\{ \begin{array}{l} SP(v_{00} \rightarrow v_{i-1,j-1}) + \Delta_{S_i T_j}, \\ SP(v_{00} \rightarrow v_{i-1,j}) + \gamma, \\ SP(v_{00} \rightarrow v_{i,j-1}) + \gamma \end{array} \right\}$$

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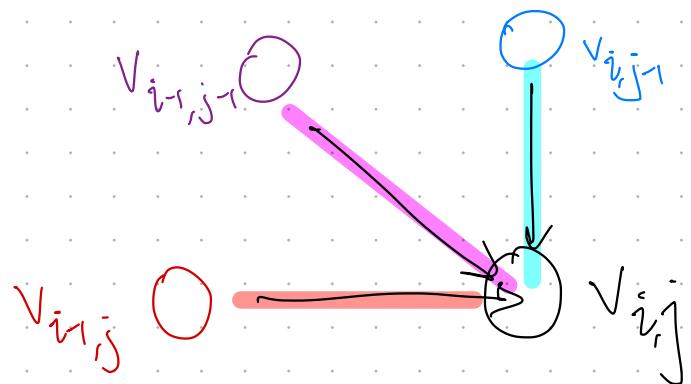
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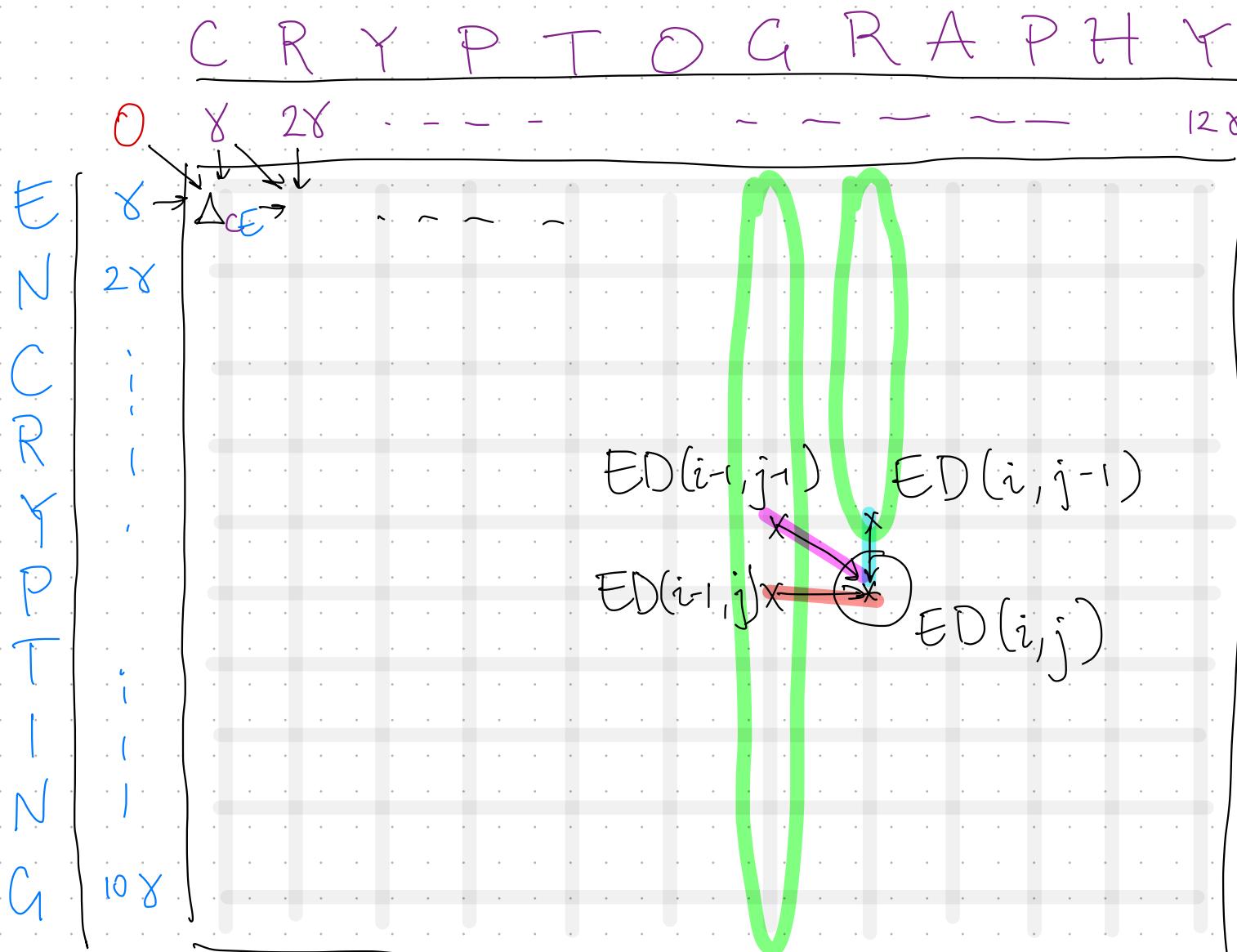
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$$SP(v_{00} \rightarrow v_{ij}) = ED[i, j]$$

□

Reducing the Space Complexity?



Each entry requires 3 probes into prior entries.

↳ From prev. column from top to bottom.

Linear Space ED.

$\text{Prev}(j) = j \cdot \gamma$ // m-entry 1D arrays

$\text{Curr}(j) = 0$

For $i=1 \rightarrow n$

$\text{Curr}(0) = i \cdot \gamma$

For $j=1 \rightarrow m$

$$\text{Curr}(j) = \min \left\{ \begin{array}{l} \text{Prev}(j-1), \text{Prev}(j), \text{Curr}(j-1) \\ + \Delta_{S_i T_j}, + \gamma, + \gamma \end{array} \right\}$$

$\text{Prev} \leftarrow \text{Curr}$

return $\text{Curr}(m)$