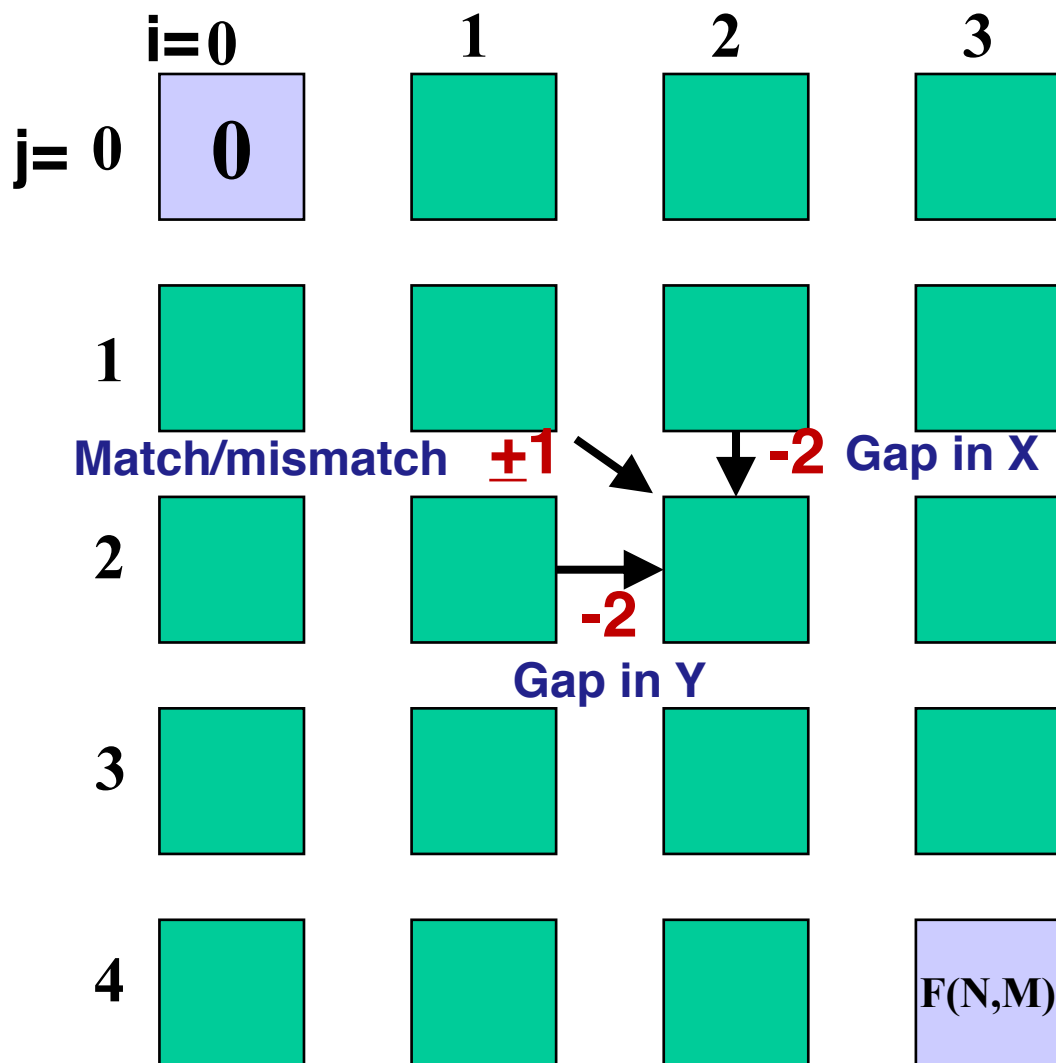


# Sequence Alignment 2

- Alignment Variants
- Affine Gap Penalty
- State Transition Trellis

# Global Alignment



$$F(i, j) = \max \begin{cases} F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

**$F(i,j)$**

**Score of best  
alignment between  
 $X_1, \dots, X_i$   
and  
 $Y_1, \dots, Y_j$**

# Global Alignment

---

	i=0	1	2	3
j= 0	0	-2	-4	-6
1	-2			
2	-4			
3	-6			
4	-8			F(N,M)

$$F(i, j) = \max \begin{cases} F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

Fill in **Boundary**  
**Conditions**

# Global Alignment

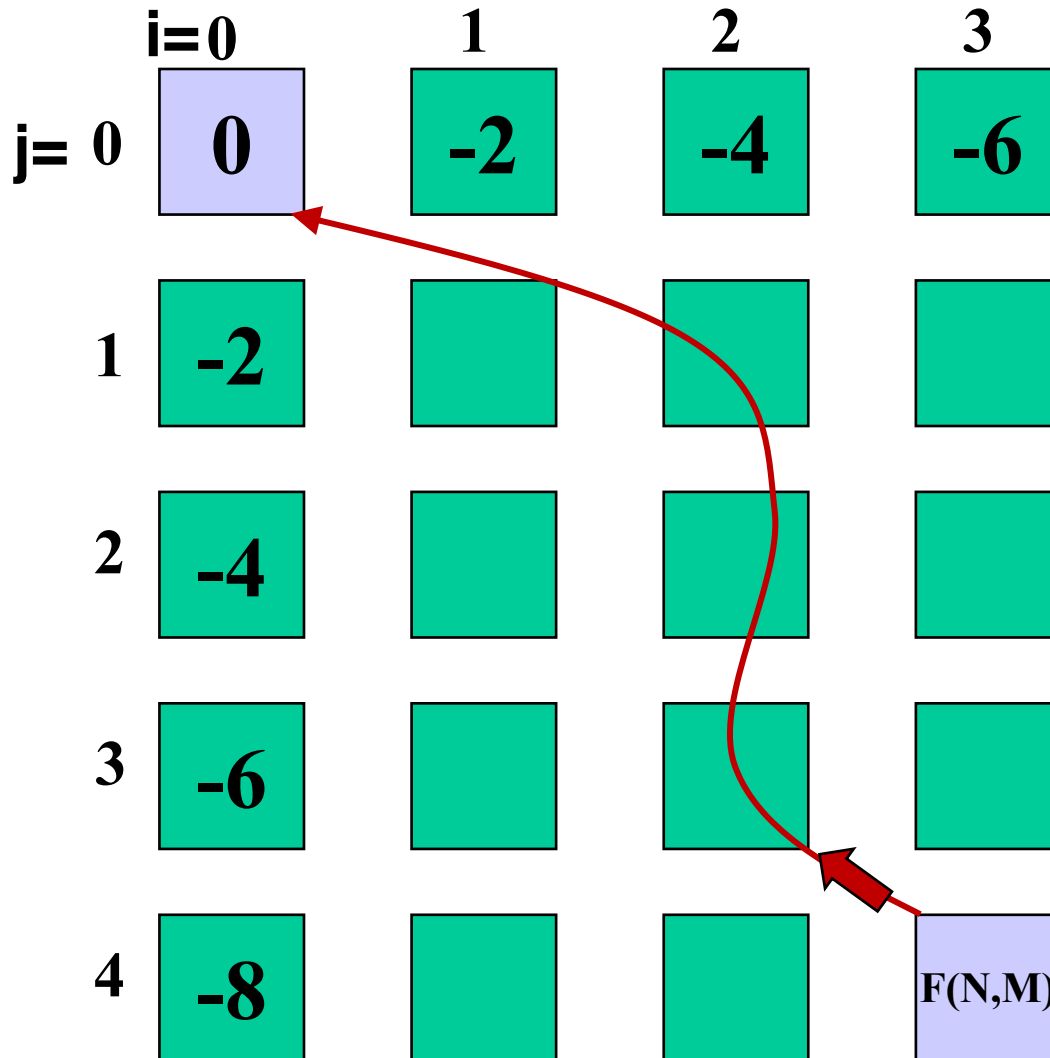
---

	i=0	1	2	3
j= 0	0	-2	-4	-6
1	-2			
2	-4			
3	-6			
4	-8			F(N,M)

$$F(i, j) = \max \begin{cases} F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

**F(N,M)**  
**Score** of best global  
 alignment

# Global Alignment



$$F(i, j) = \max \begin{cases} F(i-1, j-1) + 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

$F(N,M)$   
**Score** of best global  
 alignment

Use **Tracebacks** to get  
 actual alignments:

Store a pointer from  $F(i,j)$   
 to the cell used

# Alignment Variants

---

- Bounded Alignments
- Semi-global Alignment
- Local Alignment

# Bounded Sequence Alignment

What if  $X$  and  $Y$  are very similar?

Assume that #gaps will be small ( $< k$ )

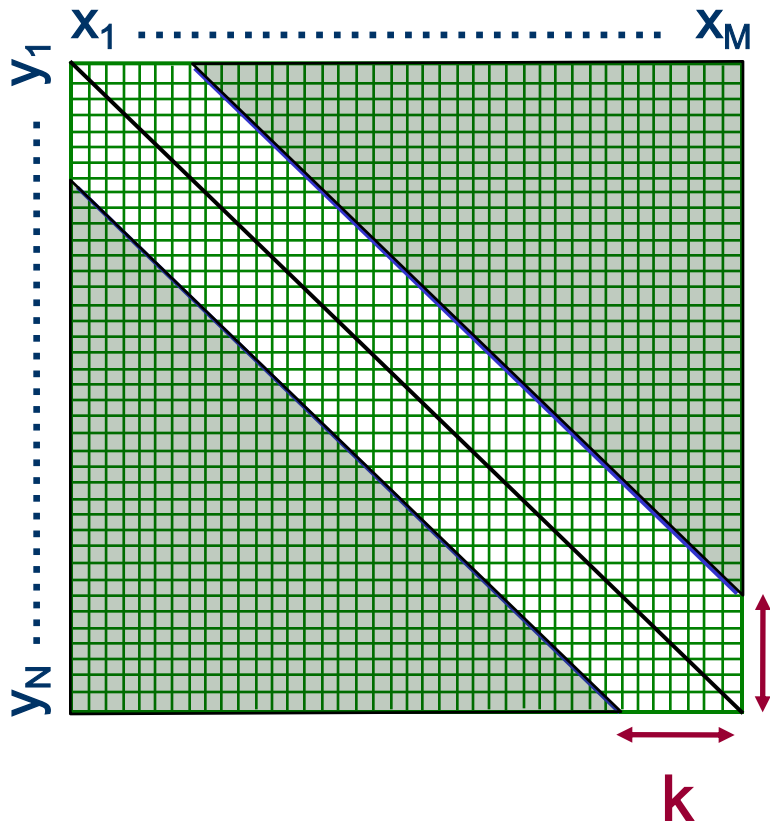
This means that if  $X_i$  matches  $Y_j$ ,  $|i-j| < k$

$X_1$	$X_2$	$X_3$	$X_4$	$X_5$	$X_6$
$Y_1$	-	-	$Y_2$	$Y_3$	$Y_4$

*Can we use this to speed up alignment?*

# Bounded Dynamic Programming

---



## Initialization:

$F(i,0)$ ,  $F(0,j)$  undefined for  $i, j > k$

## Iteration Rule:

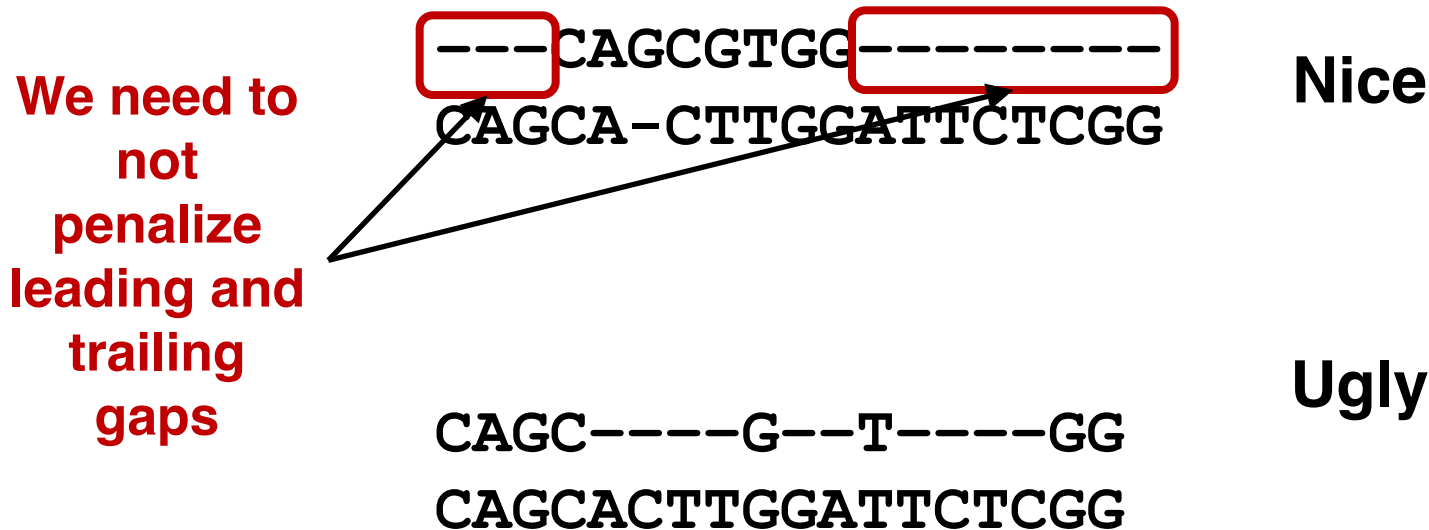
$$F(i, j) = \max \begin{cases} F(i-1, j-1) + s(x_i, y_j) \\ F(i, j-1) - d, \text{ if } j > i - k \\ F(i-1, j) - d, \text{ if } j < i + k \end{cases}$$



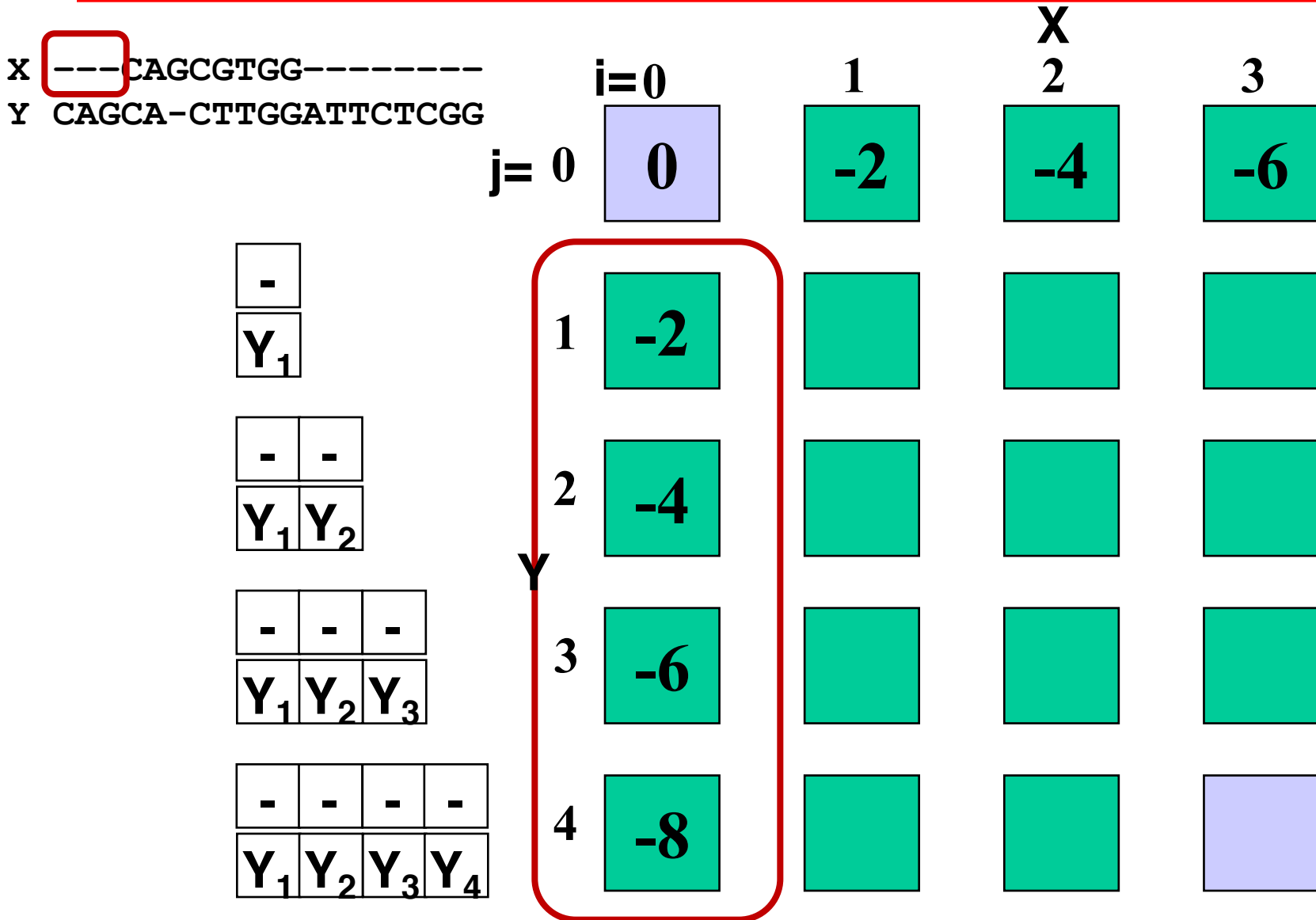
# Semi-Global Alignment

---

- Sometimes an entire sequence (X) is embedded in another sequence (Y)
- e.g. complete protein domain to a gene



# Semi-global Alignment



# Semi-global Alignment

X ---CAGCGTGG-----  
 Y CAGCA-CTTGGATTCTCGG

	i=0	1	2	3
j= 0	0	-2	-4	-6

-  
Y<sub>1</sub>

1	0			
---	---	--	--	--

- -  
Y<sub>1</sub> Y<sub>2</sub>

2	0			
---	---	--	--	--

- - -  
Y<sub>1</sub> Y<sub>2</sub> Y<sub>3</sub>

3	0			
---	---	--	--	--

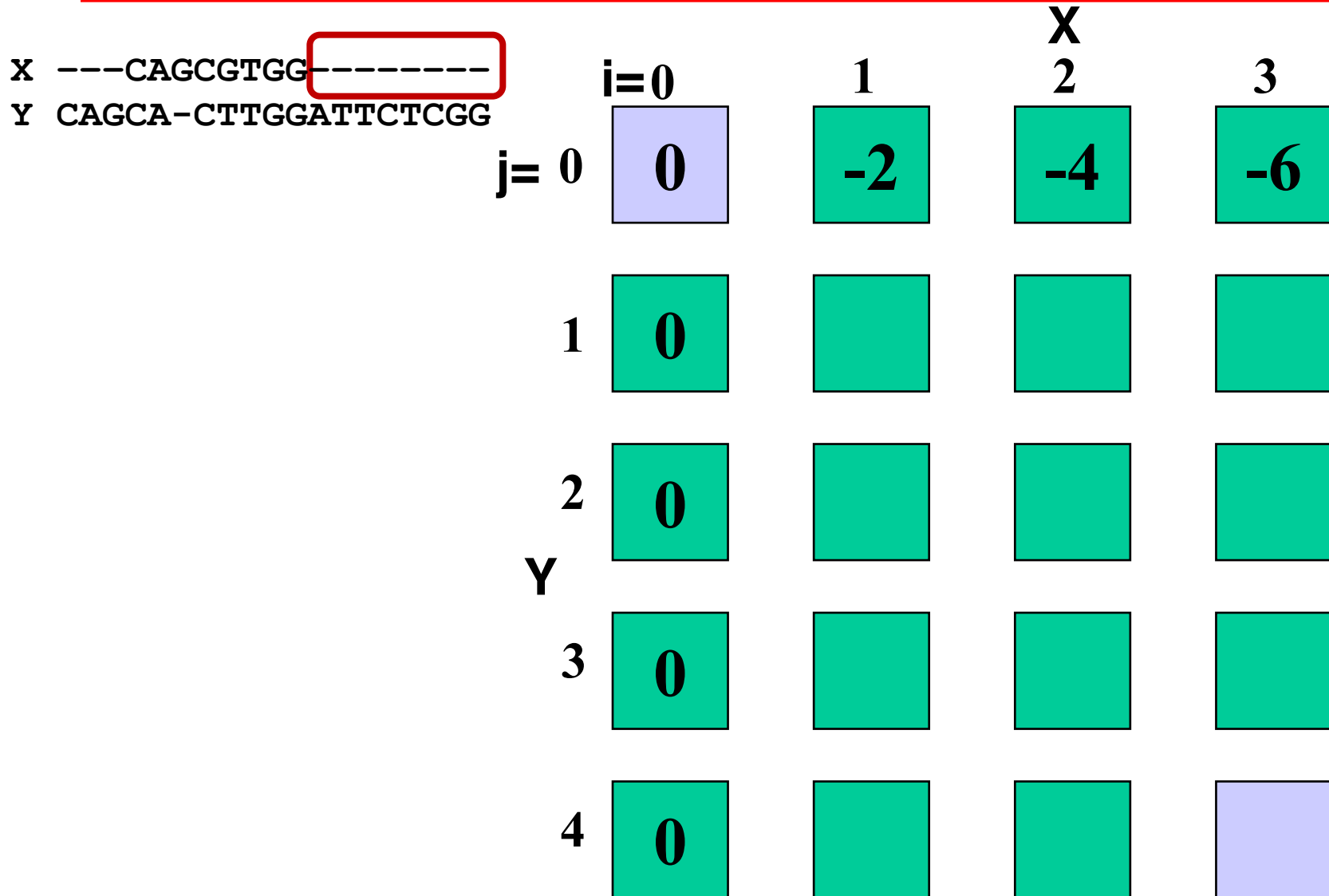
- - - -  
Y<sub>1</sub> Y<sub>2</sub> Y<sub>3</sub> Y<sub>4</sub>

4	0			
---	---	--	--	--

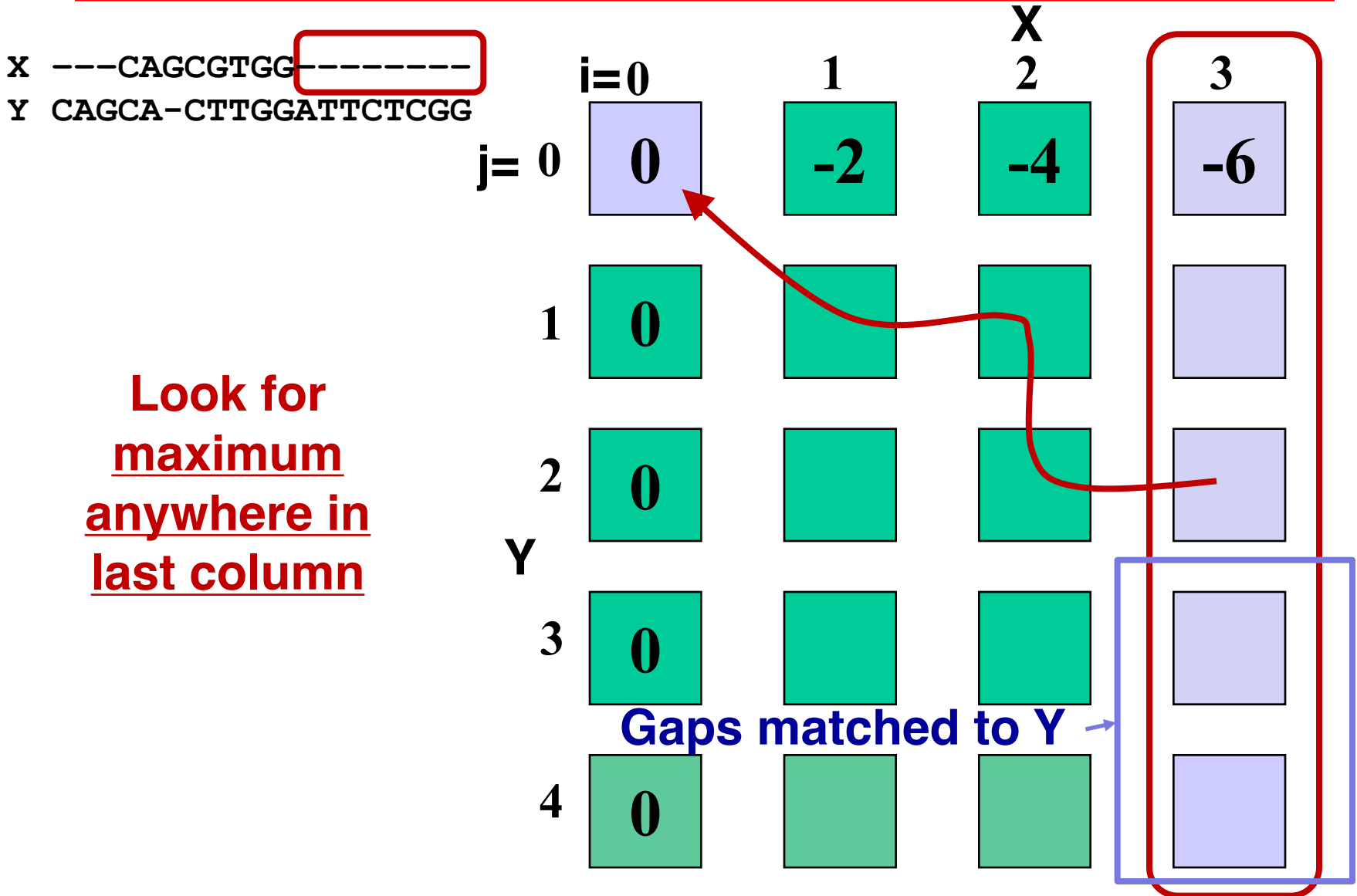
**First  
Row  
all  
zeros**

# Semi-global Alignment

---



# Semi-global Alignment



Y ---CAGCGTGG-----

1

# X 2

3

**j= 0**

0

1

2

3

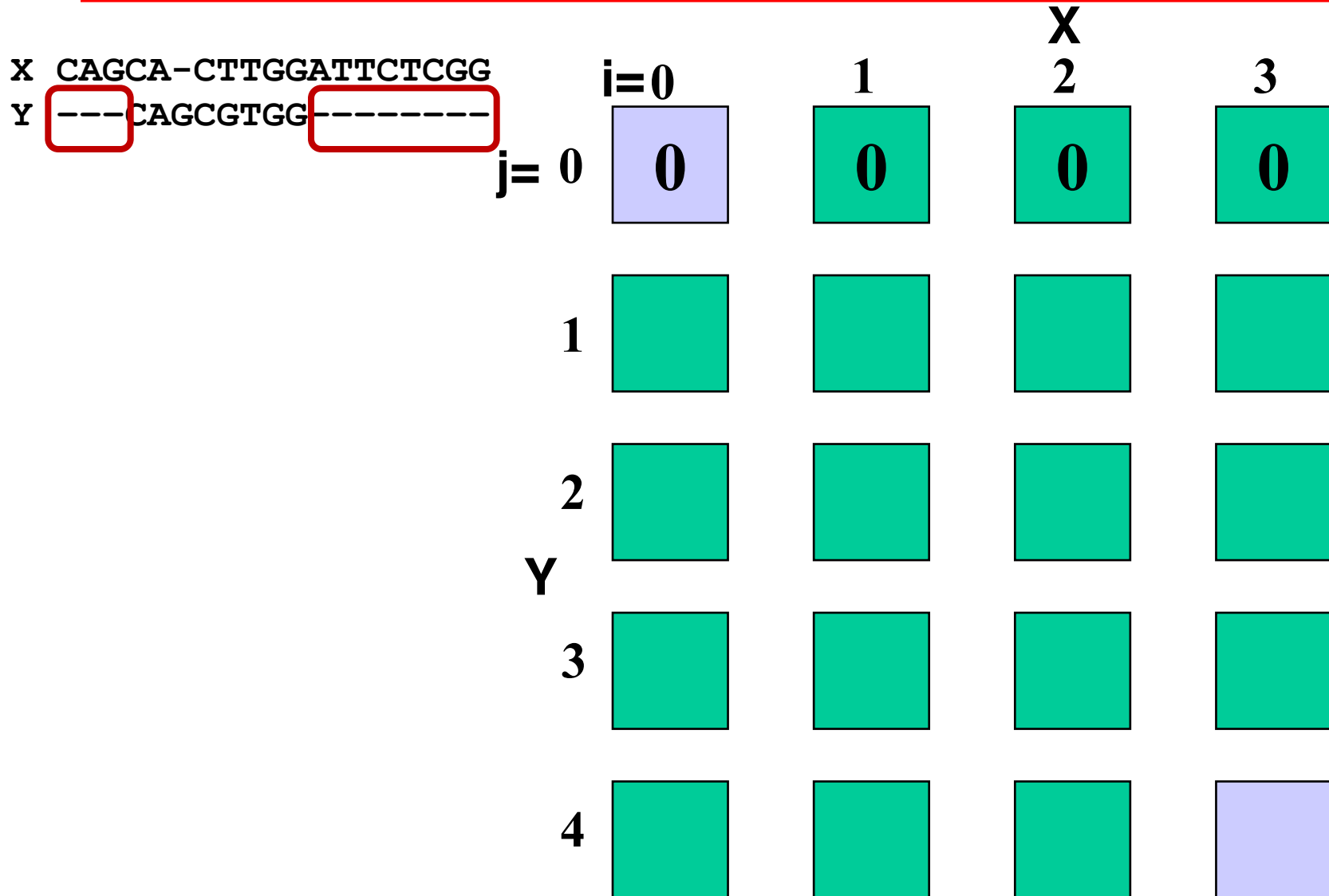
4

# Y

# X 2

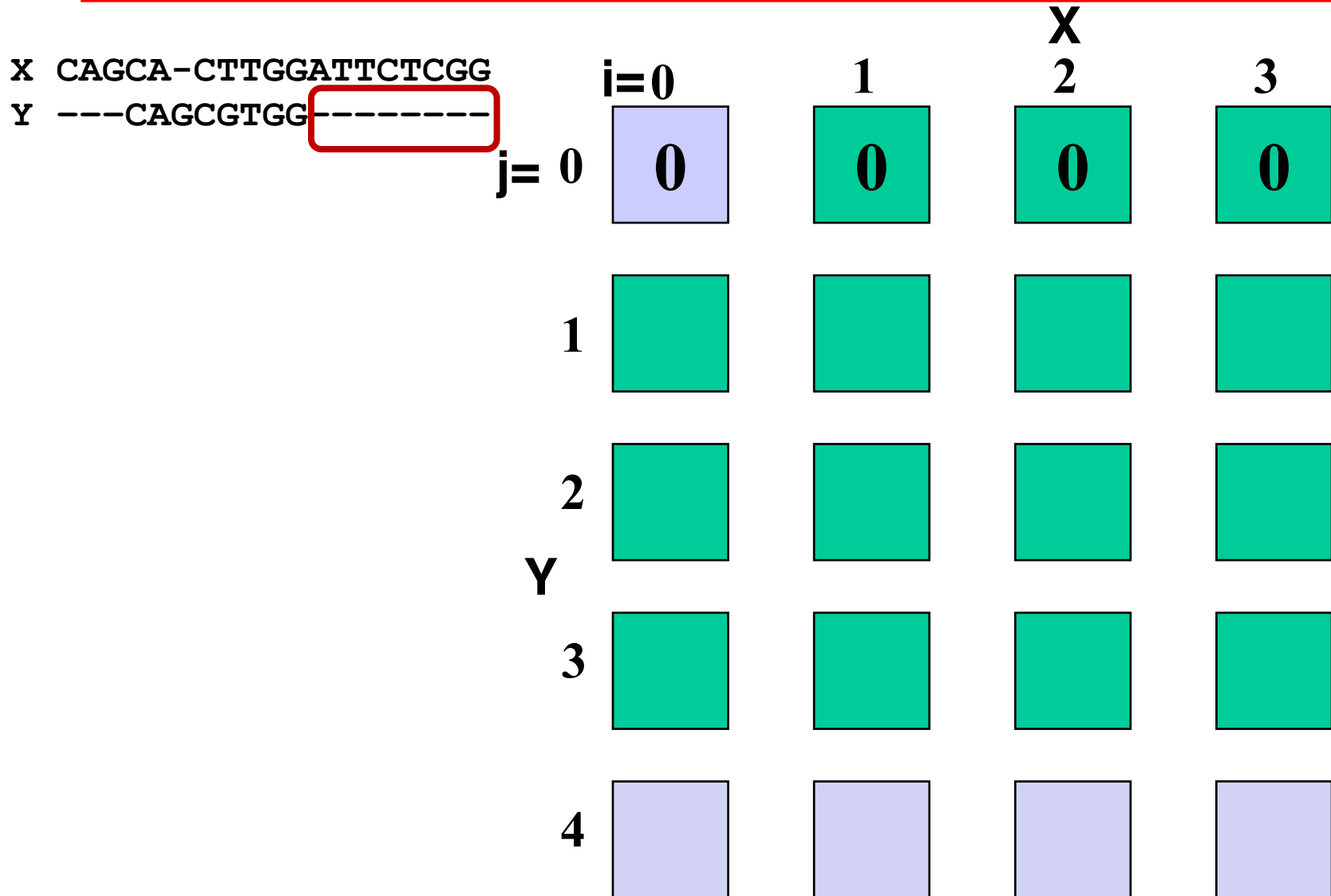
# Semi-global Alignment

---



# Semi-global Alignment

---





# Alignment Variations

---

- Summary of end space scoring procedures:

<b>Place where spaces are not penalized for</b>	<b>Action</b>
<b>Before 1<sup>st</sup> sequence</b>	<b>Initialize 1<sup>st</sup> column with zeros</b>
<b>After 1<sup>st</sup> sequence</b>	<b>Look for max in last column</b>
<b>Before 2<sup>nd</sup> sequence</b>	<b>Initialize 1<sup>st</sup> row with zeros</b>
<b>After 2<sup>nd</sup> sequence</b>	<b>Look for max in last row</b>

# Semi-global Alignment

---

**Allow  
gaps at  
either end  
of each  
sequence**

**But this  
is still a  
global  
alignment**

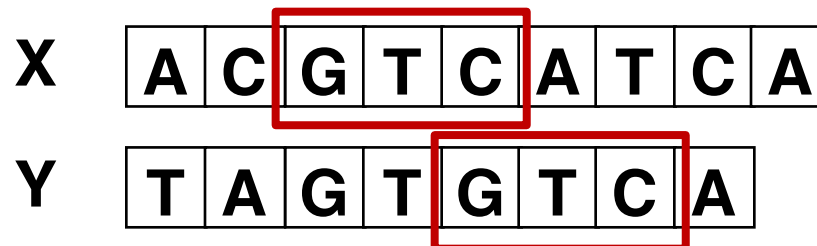
		X			
		i=0	1	2	3
Y	j= 0	0	0	0	0
	1	0			
	2	0			
	3	0			
	4	0			

# Local Alignment

---

## Smith-Waterman Algorithm

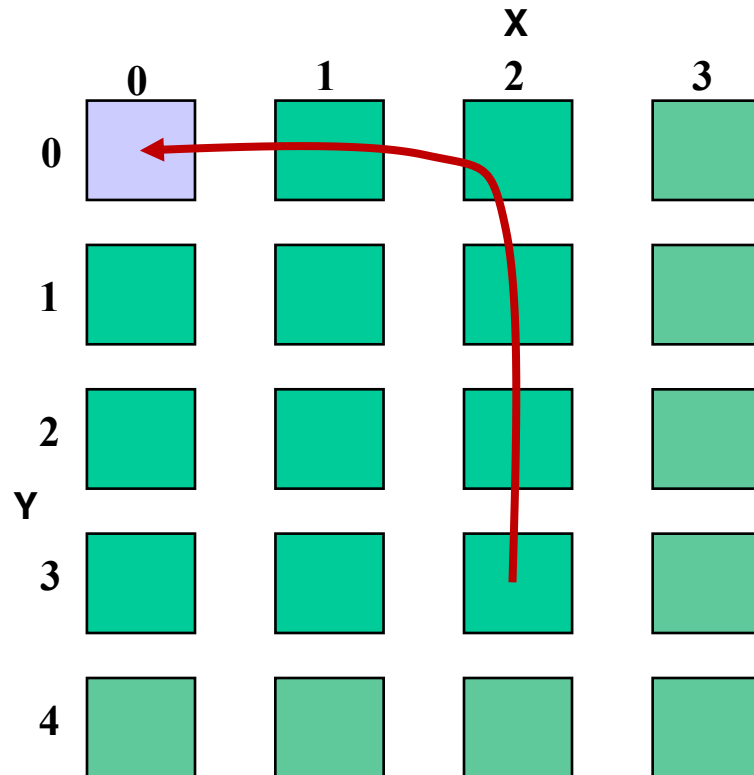
- Sometimes we just want an alignment between **part of X** and **part of Y**
- e.g. conserved protein domain between two complete genes
- This means finding the highest scoring alignment of any subsequences of X & Y



# A Slightly Easier Problem

---

X	A	C	G	T	C	A	T	C	A
Y	T	T	C	T	G	T	C	A	



**What if we want the  
highest score alignment  
of any two substrings  
that start at  $X_0, Y_0$**

# A Slightly Easier Problem

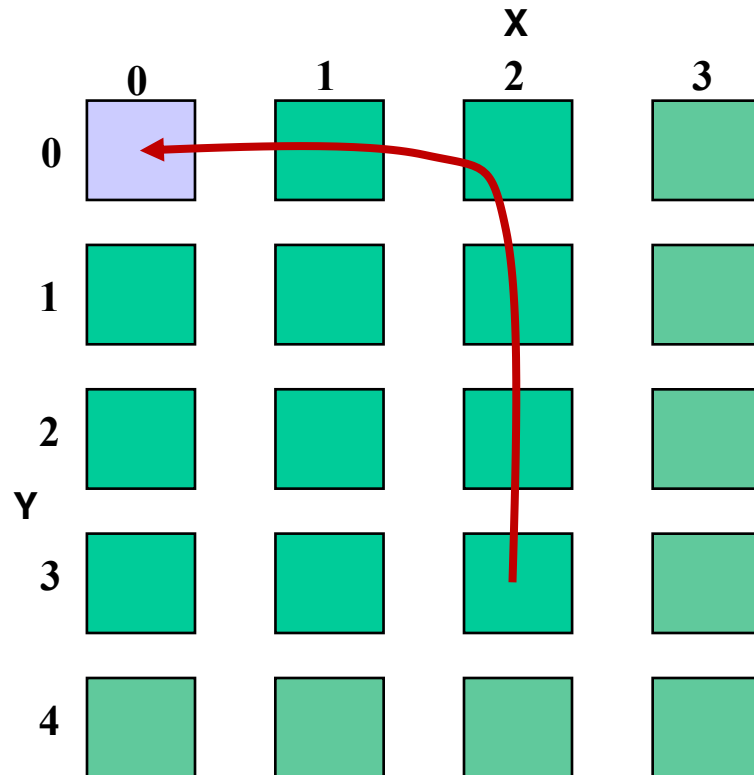
---

X    

A	C	G	T	C	A	T	C	A
---	---	---	---	---	---	---	---	---

Y    

T	T	C	T	G	T	C	A
---	---	---	---	---	---	---	---

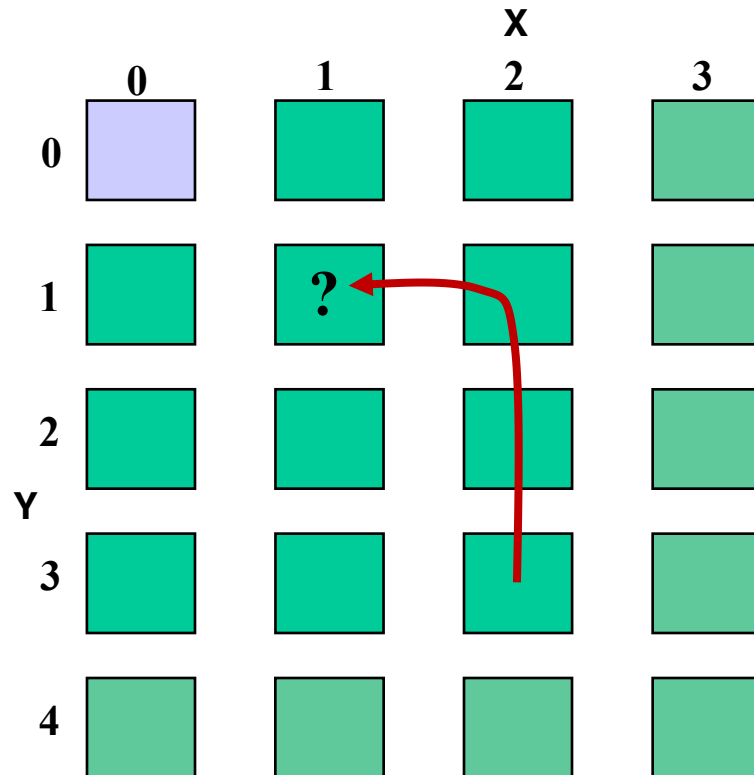


A	-	C	-	G	T	C
T	T	C	T	G	T	C
-1	-2	1	-2	1	1	1

# Smith-Waterman Algorithm

---

X    A C G T C A T C A  
Y    T T C T G T C A



Scoring Matrix:

A	-	C	-	G	T	C
T	T	C	T	G	T	C
-1	-2	1	-2	1	1	1

# Smith-Waterman Algorithm

---

X	A	C	G	T	C	A	T	C	A
Y	T	T	C	T	G	T	C	A	

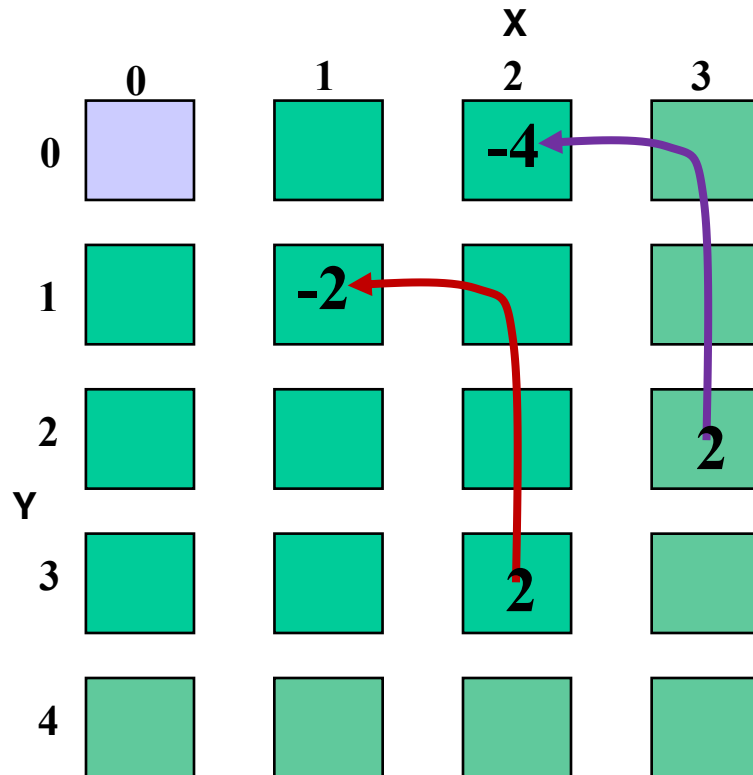
		X			
	0	1	2	3	
0					
1		$\leq 0$			
2					
3					
4					

**Stop the alignment  
when what came  
before had more  
bad than good**

# Smith-Waterman Algorithm

---

But imagine this hypothetical scenario  
Is the blue alignment really just as good as the red?



*No, because it started at a lower negative to reach 2!*

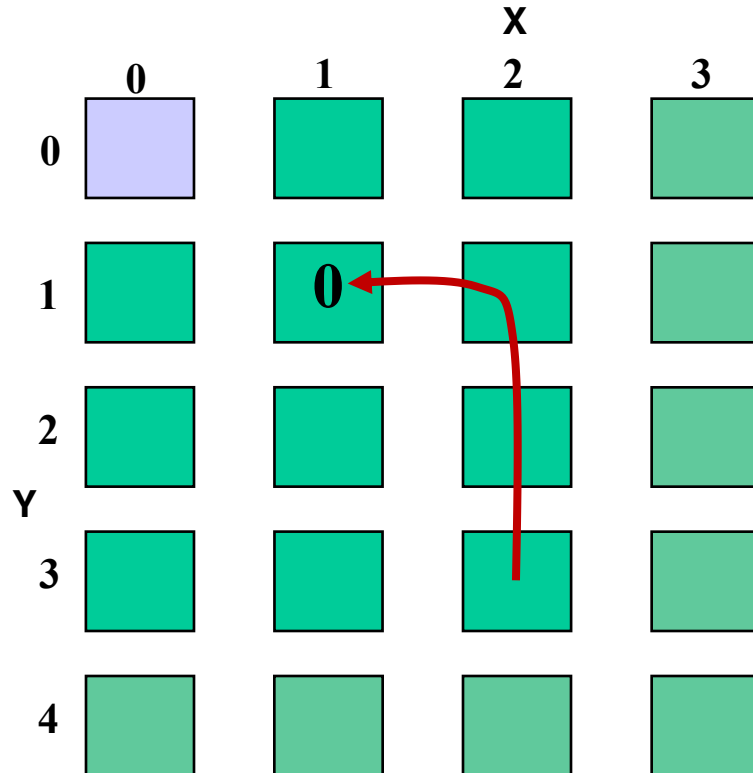


# Smith-Waterman Algorithm

---

$$F(i, j) = \max \begin{cases} \boxed{0} \\ F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

During recursion,  
rather than going  $< 0$ ,  
we reset  $F$  to zero!



Start alignment at  
**highest  $F(i,j)$  anywhere**  
in matrix – traceback  
to **first zero**

# Smith-Waterman Algorithm

---

$$F(i, j) = \max \left\{ \begin{array}{l} 0 \\ F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{array} \right.$$

During recursion,  
rather than going  $< 0$ ,  
we reset  $F$  to zero!

	0	1	2	3
0	0	0	0	0
1	0	0		
2	0			
3	0			
4	0			

Initialize first row and  
column to zeros

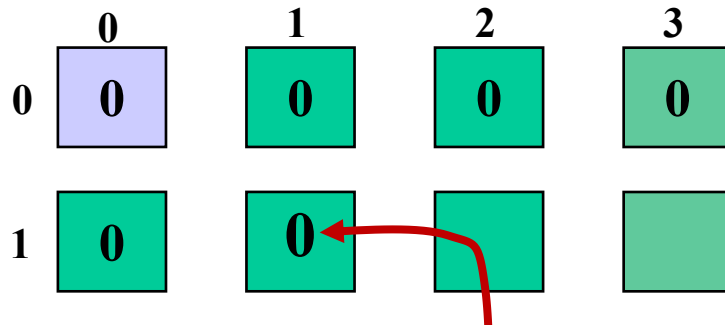
Start alignment at  
**highest  $F(i,j)$  anywhere**  
in matrix – traceback  
to **first zero**

# Smith-Waterman Algorithm

---

$$F(i, j) = \max \left\{ \begin{array}{l} \boxed{0} \\ F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{array} \right.$$

**During recursion,  
rather than going  $< 0$ ,  
we reset F to zero!**



**Initialize first row and  
column to zeros**

**Smith-Waterman is guaranteed to find the highest  
scoring alignment between *any* substring of X and *any*  
substring of Y\***

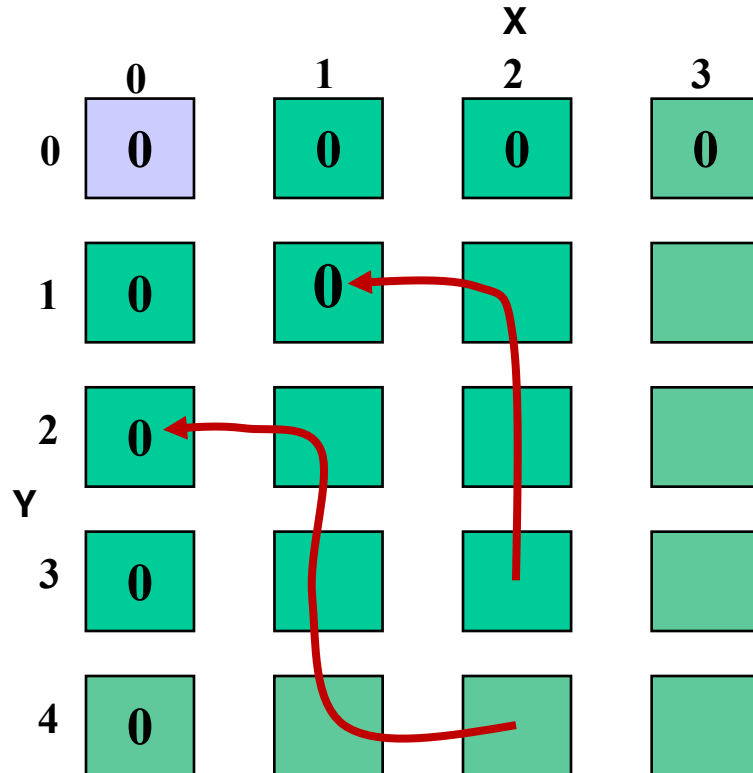
**(\*Must have  $E[\text{score of random seqs}] < 0$ )**

# Smith-Waterman Algorithm

---

$$F(i, j) = \max \begin{cases} \boxed{0} \\ F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

**During recursion,  
rather than going  $< 0$ ,  
we reset F to zero!**



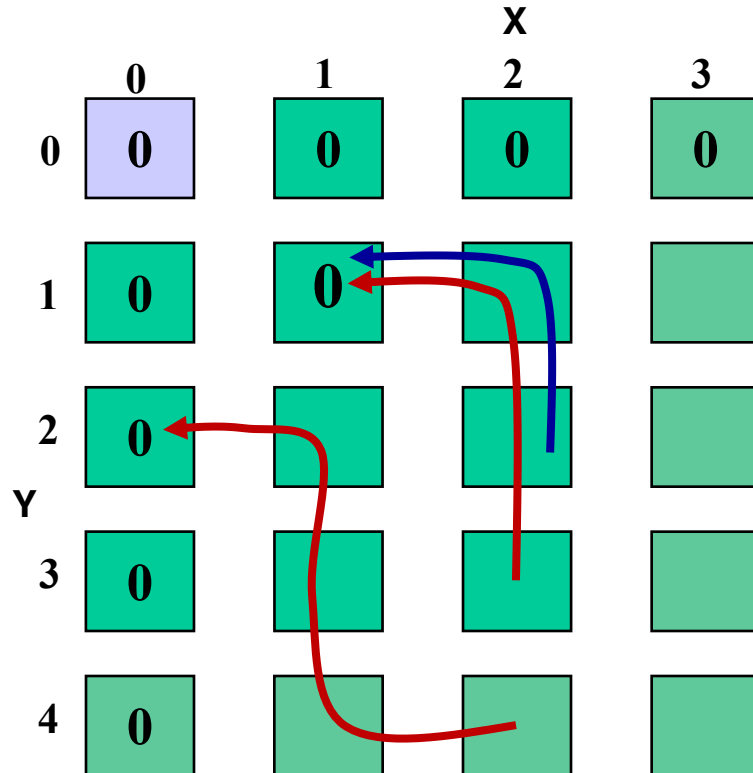
**What about the next  
highest scoring  
subsequence  
alignment?**

# Smith-Waterman Algorithm

---

$$F(i, j) = \max \begin{cases} \boxed{0} \\ F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

**During recursion,  
rather than going  $< 0$ ,  
we reset F to zero!**



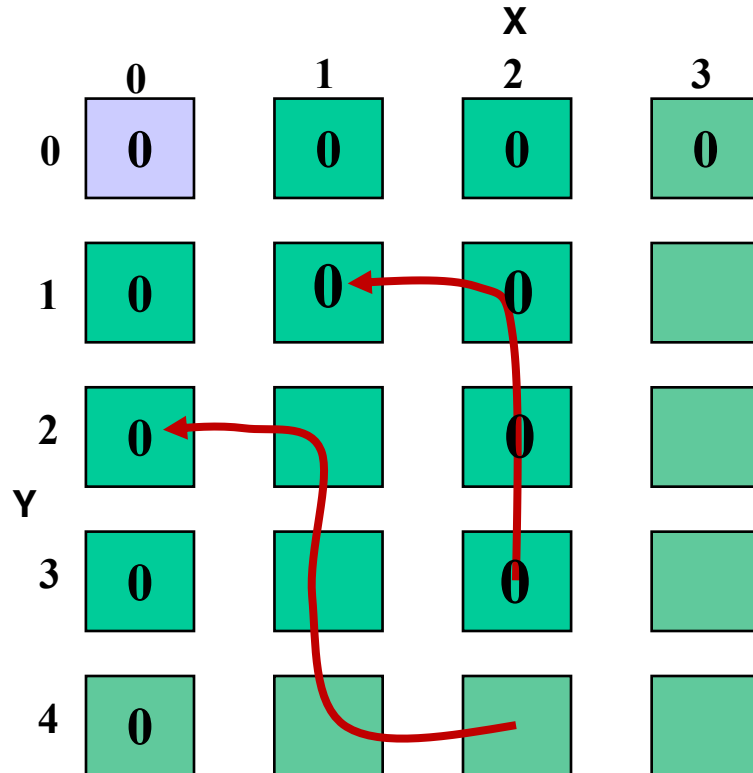
**But how do we avoid  
this trivial solution?**

# Smith-Waterman Algorithm

---

$$F(i, j) = \max \begin{cases} \boxed{0} \\ F(i-1, j-1) \pm 1 \\ F(i-1, j) - 2 \\ F(i, j-1) - 2 \end{cases}$$

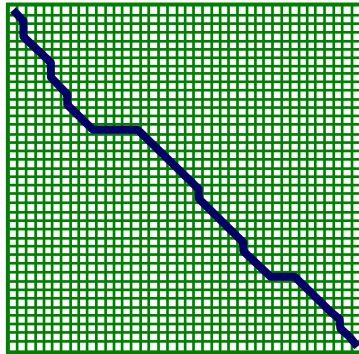
**During recursion,  
rather than going  $< 0$ ,  
we reset F to zero!**



**But how do we avoid  
this trivial solution?**

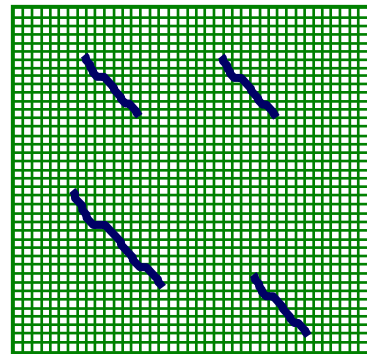
**Zero out found  
alignments**

**Global**



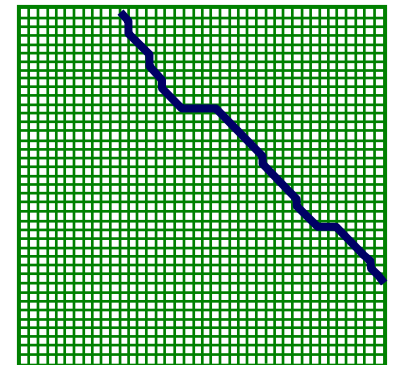
**Complete alignment**

**Local**



**Stretches of similarity**

**Semi-global**



**No end-gap penalty**

**Initialization  
to zero**

**Top left**

**Top row/left col.**

**Top row or  
left column**

**Iteration: max**

$$\begin{aligned} &F(i-1, j) - d \quad \star \\ &F(i, j-1) - d \\ &F(i-1, j-1) + s(x_i, y_j) \end{aligned}$$

$$\begin{aligned} &0 \\ &F(i-1, j) - d \\ &F(i, j-1) - d \\ &F(i-1, j-1) + s(x_i, y_j) \end{aligned}$$

$$\begin{aligned} &F(i-1, j) - d \\ &F(i, j-1) - d \\ &F(i-1, j-1) + s(x_i, y_j) \end{aligned}$$

**Termination**

**Bottom right**

**Anywhere**

**Bottom row  
or right column**

# Exercise for You

---

- Often gaps come in bunches
- The probability of getting 10 gaps in a row is higher than 10 different 1 gap events
- Solution: Affine Gap Penalty
- $g(n)$  is penalty for gap of length  $n$ :

$$g(n) = \underset{\substack{\nearrow \\ \text{Gap-open penalty (d)}}}{(-d)} - (n-1) \underset{\substack{\nearrow \\ \text{Gap-extension penalty (e)}}}{e}$$

**What is the new update rule for  $F(i,j)$ ?**



# Key Idea

---

To compute optimal alignment, at position  $(i, j)$ , need to “remember” **both best score if gap is open AND if gap is not open**

E.g. Two scenarios for  $X_i$  aligned to gap

$x_{i-1}$  aligns to  $y_{j-1}$

A	A	$x_{i-1}$	$x_i$
A	A	$y_{i-1}$	-



**Score with  $d$**   
(gap open)

$x_{i-1}$  aligns to gap

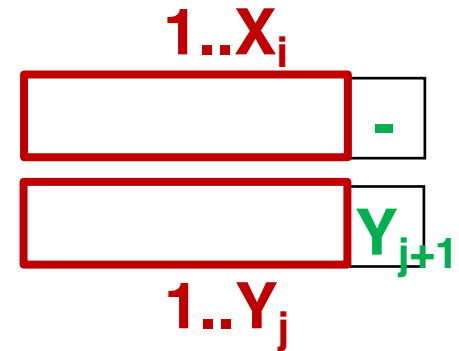
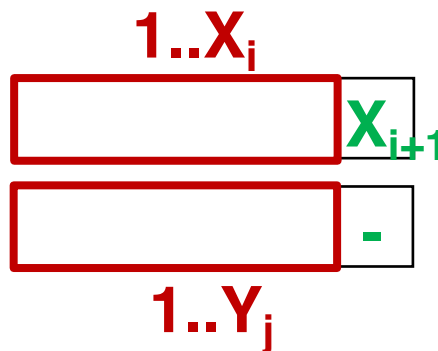
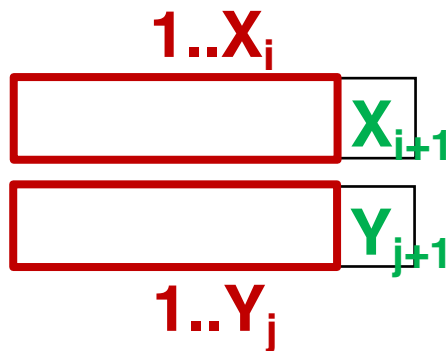
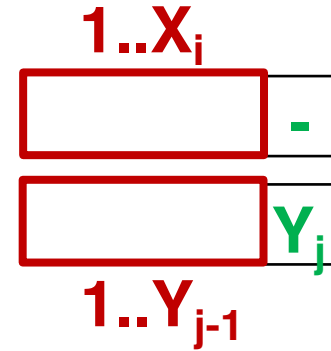
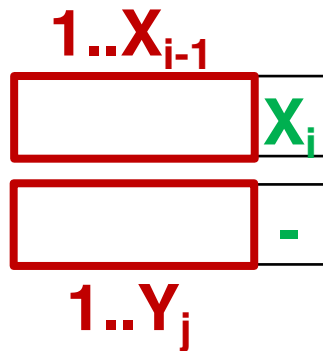
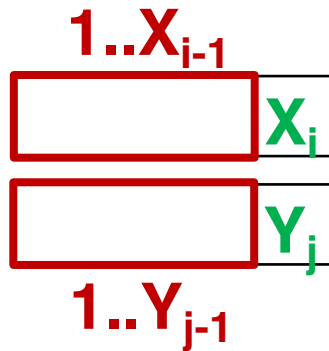
A	A	$x_{i-1}$	$x_i$
A	$y_{i-1}$	-	-



**Score with  $e$**   
(gap extension)

# Getting to $(i+1, j+1)$ from $(i, j)$

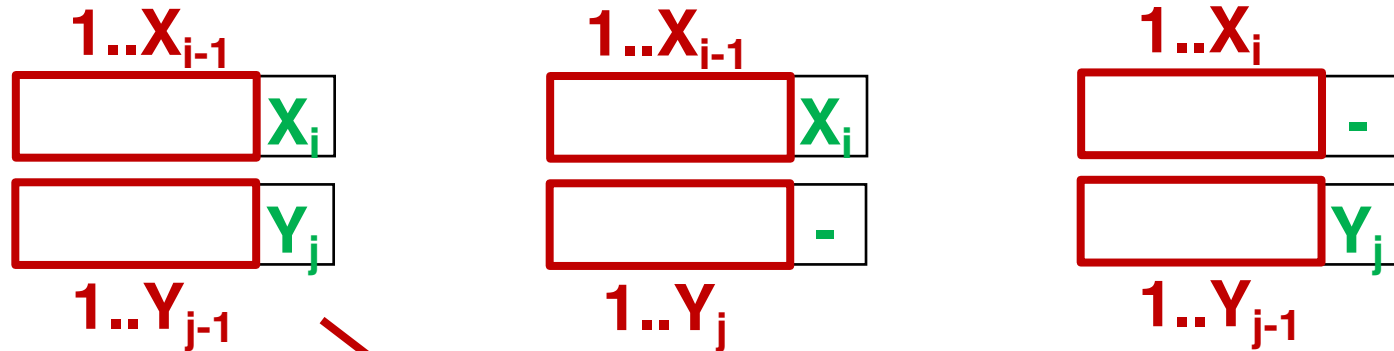
Still three ways that an alignment of  $(i, j)$  or  $(i+1, j+1)$  can end



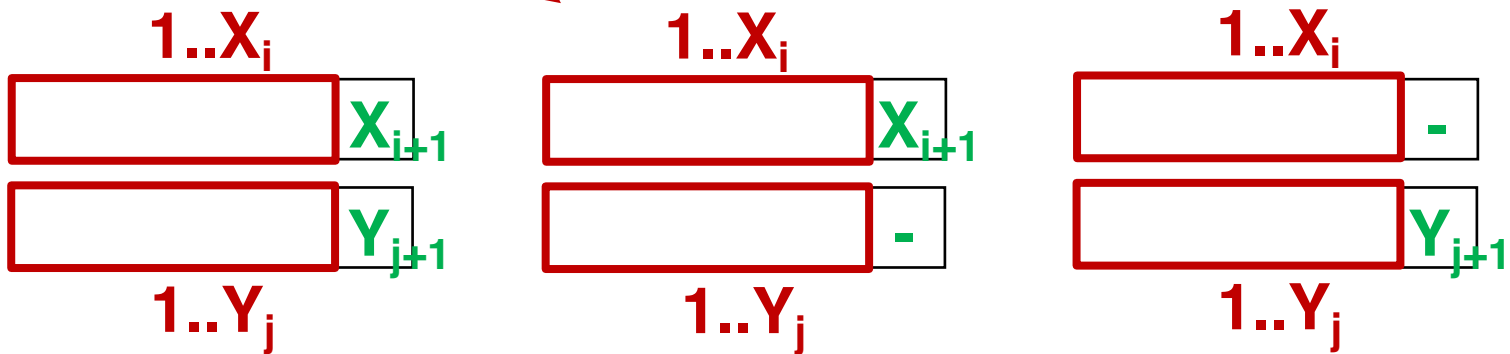
*But now it matters which way  $(i, j)$  ended to score  $(i+1, j+1)$*

# Getting to $(i+1, j+1)$ from $(i, j)$

Still three ways that an alignment of  $(i, j)$  or  $(i+1, j+1)$  can end



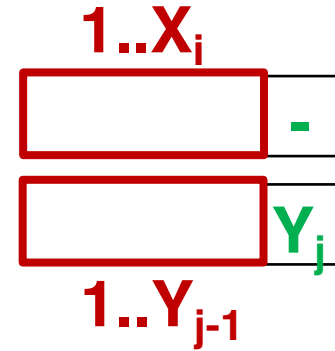
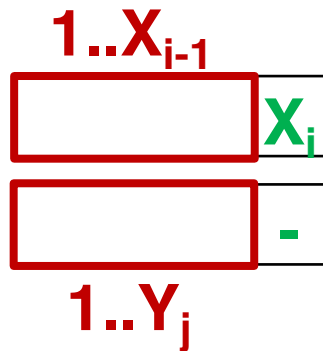
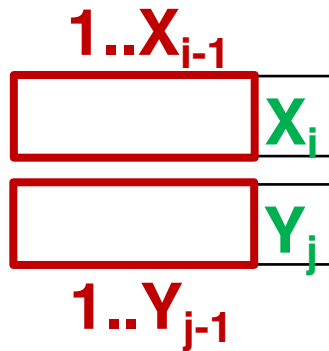
How do we score this?



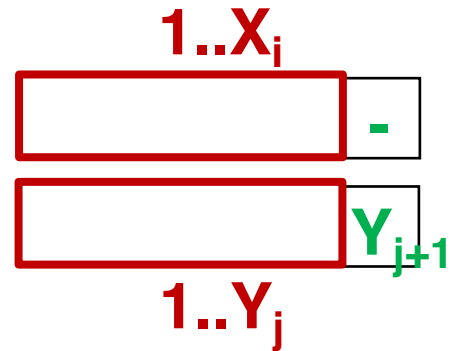
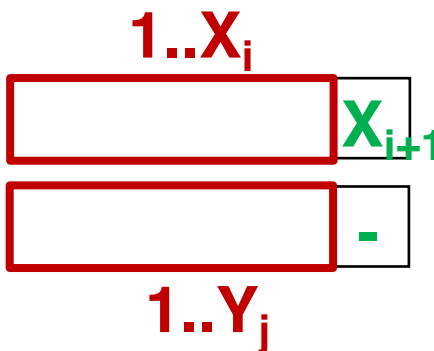
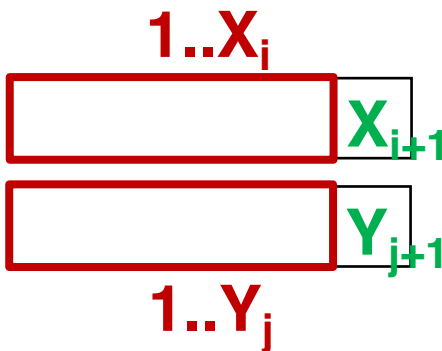
*But now it matters which way  $(i, j)$  ended to score  $(i+1, j+1)$*

# Getting to $(i+1, j+1)$ from $(i, j)$

Still three ways that an alignment of  $(i, j)$  or  $(i+1, j+1)$  can end



How do we score this?



*But now it matters which way  $(i, j)$  ended to score  $(i+1, j+1)$*

# Update Rule

---

## Define 3 Update Quantities

$M(i,j)$  – best score up to  $(i,j)$  with  $x_i$  aligned to  $y_j$

$I_x(i,j)$  – best score up to  $(i,j)$  with  $x_i$  aligned to gap

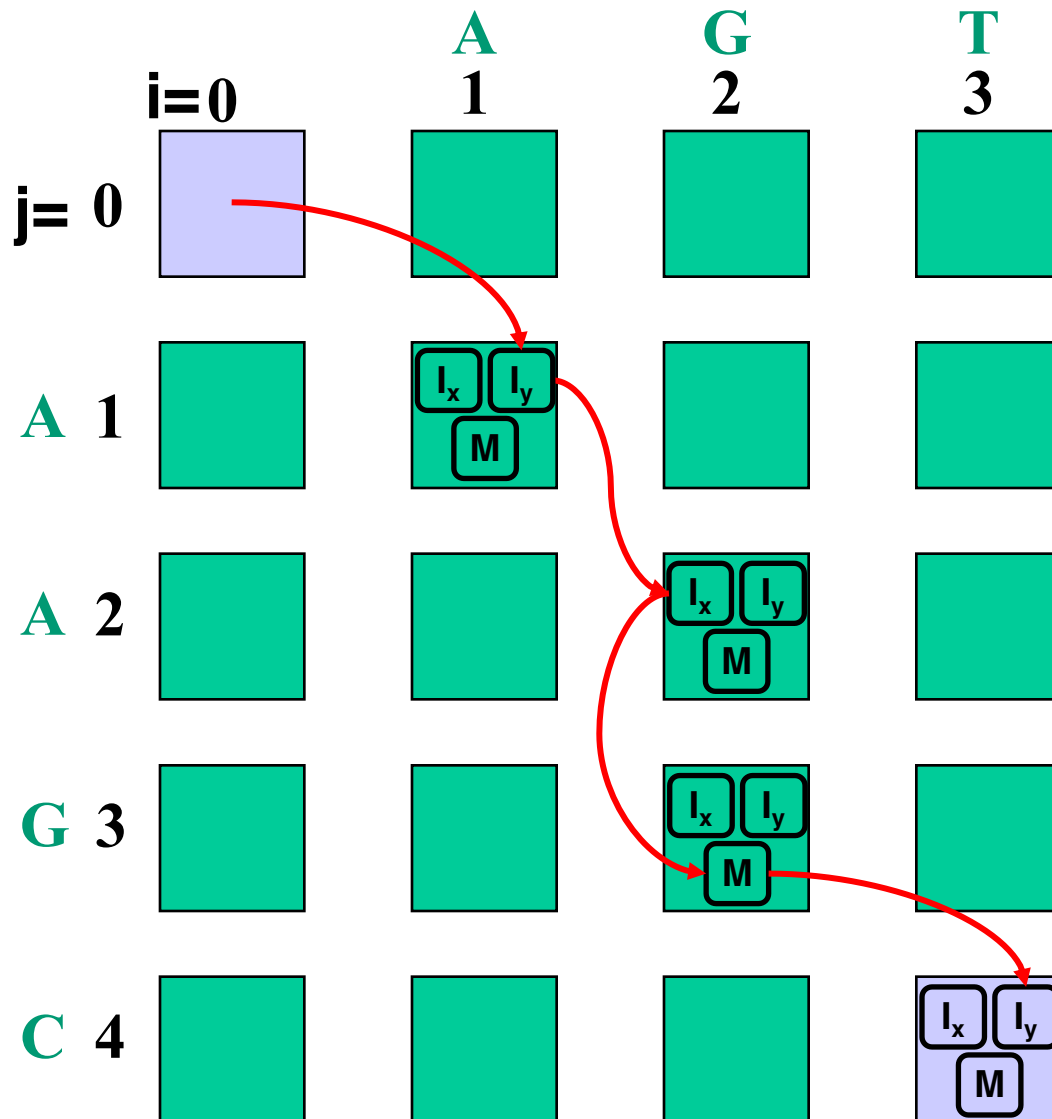
$I_y(i,j)$  – best score up to  $(i,j)$  with  $y_j$  aligned to gap

$$M(i, j) = \max \begin{cases} M(i-1, j-1) + s(x_i, y_j) \\ I_x(i-1, j-1) + s(x_i, y_j) \\ I_y(i-1, j-1) + s(x_i, y_j) \end{cases}$$

$$I_x(i, j) = \max \begin{cases} M(i-1, j) - d \\ I_x(i-1, j) - e \end{cases}$$

$$I_y(i, j) = \max \begin{cases} M(i, j-1) - d \\ I_y(i, j-1) - e \end{cases}$$

# Expanding the F Matrix



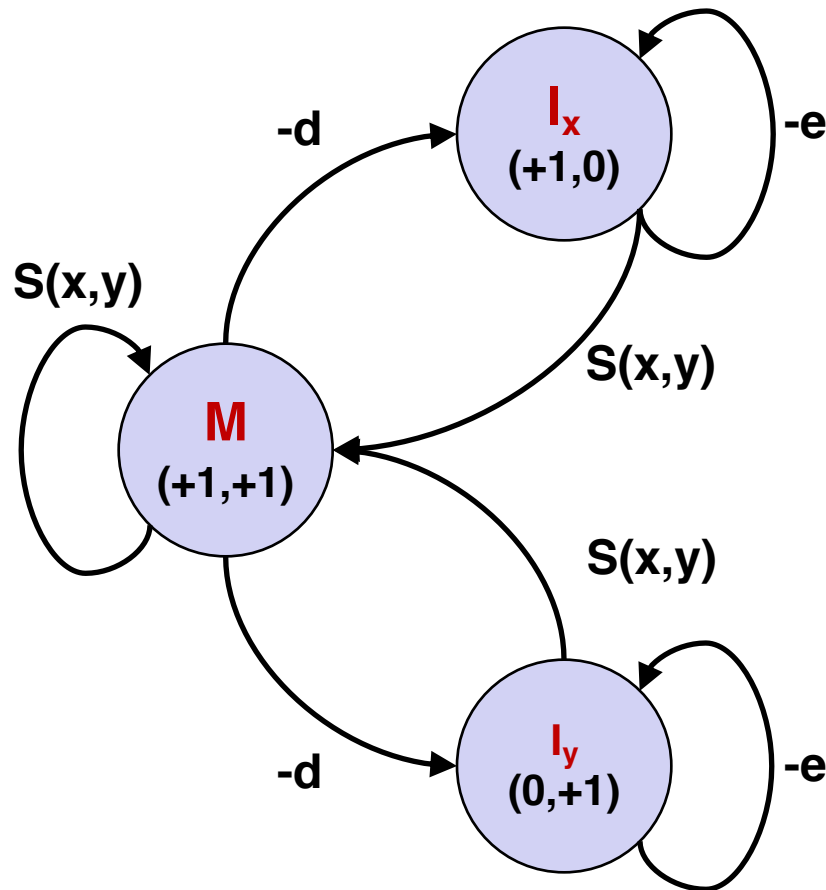
Instead of storing a single value  $F(i,j)$  in each square

We need to store three values

These can be considered 3 *states* at each  $(i,j)$

# Finite State Machine

---



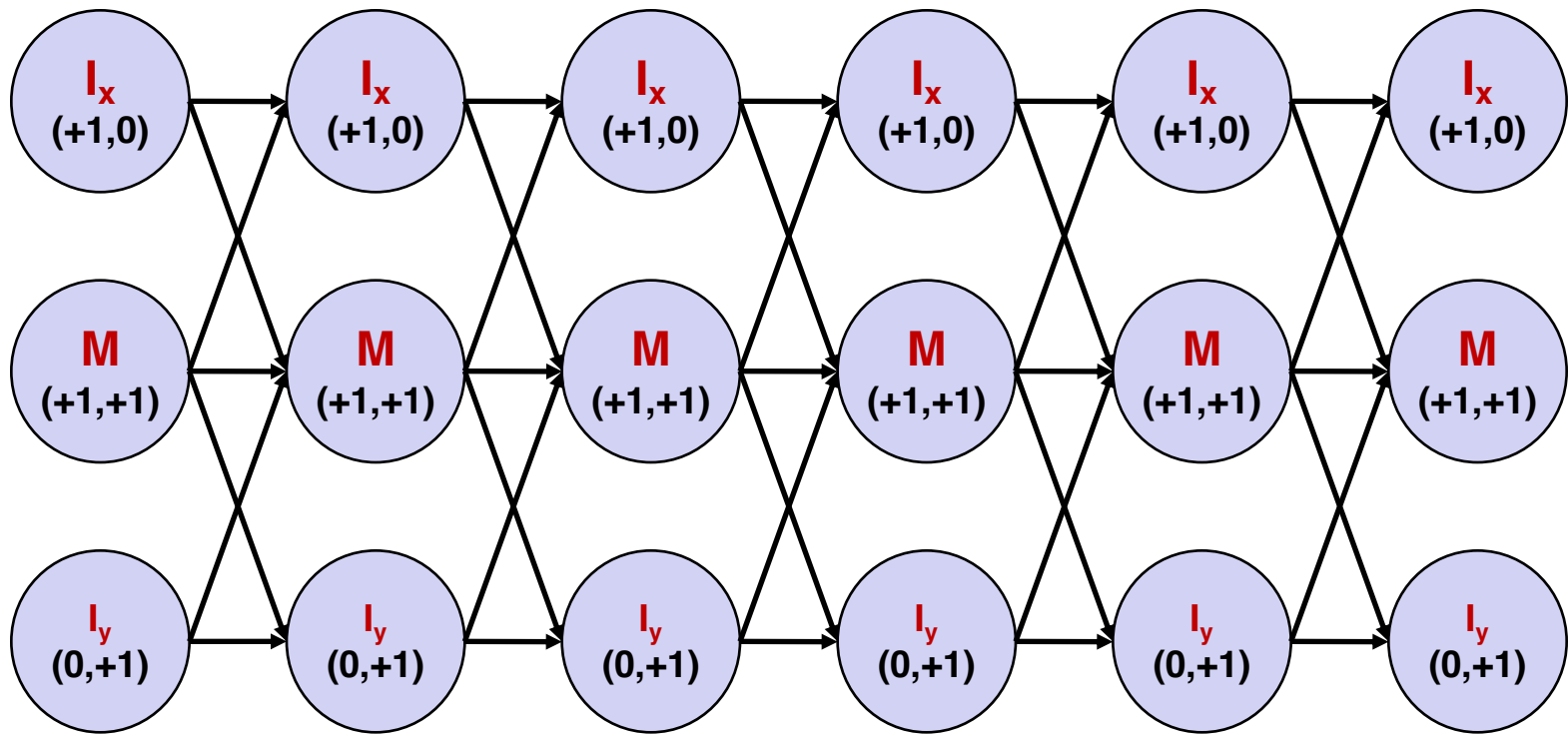
**Any given alignment  
is a series of state  
transitions through  
this state machine**

x	A	T	G	C	-	T
y	A	-	-	C	A	T

**And we can score an  
alignment by scoring  
transitions**

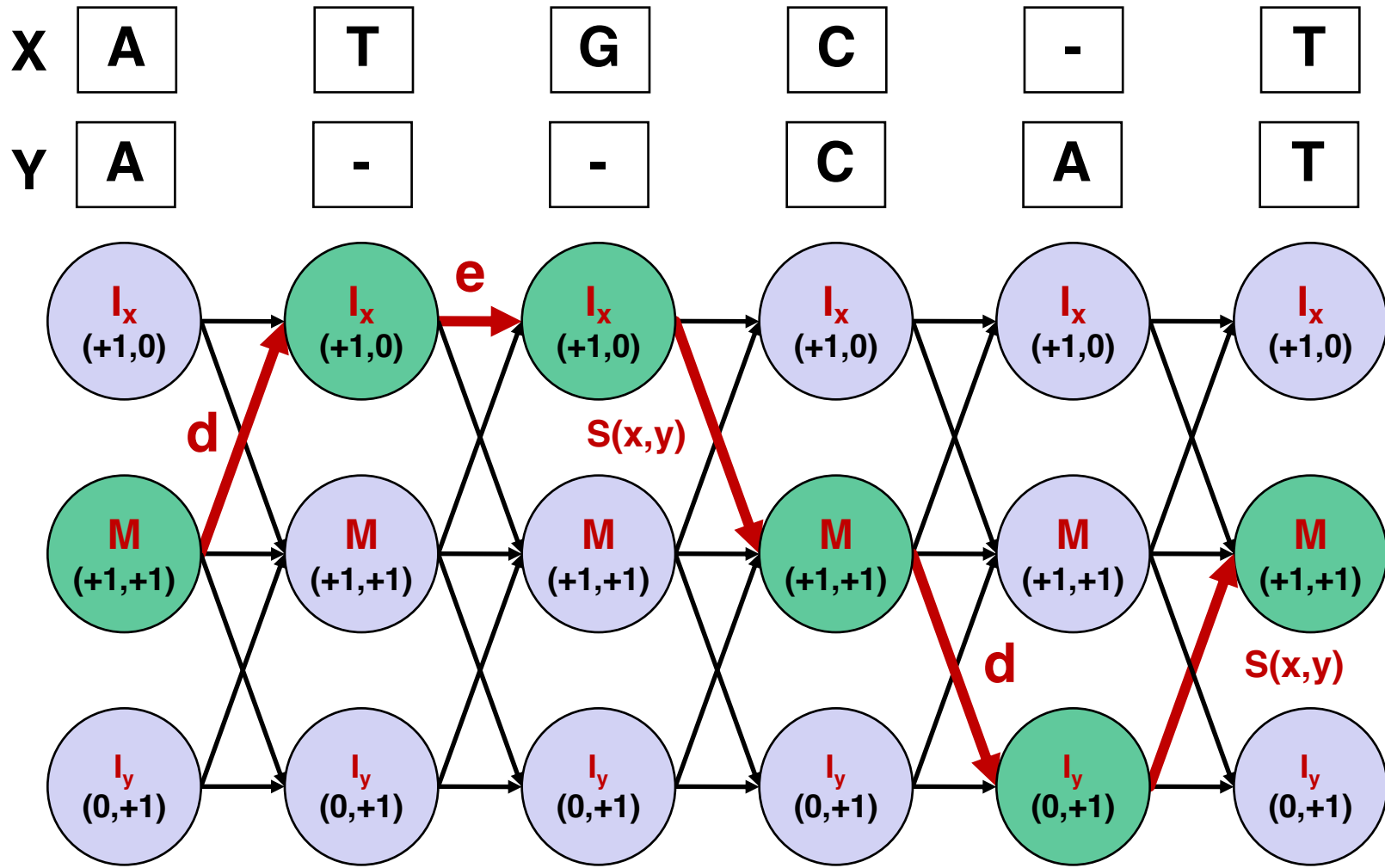
# State Transition Trellis

---





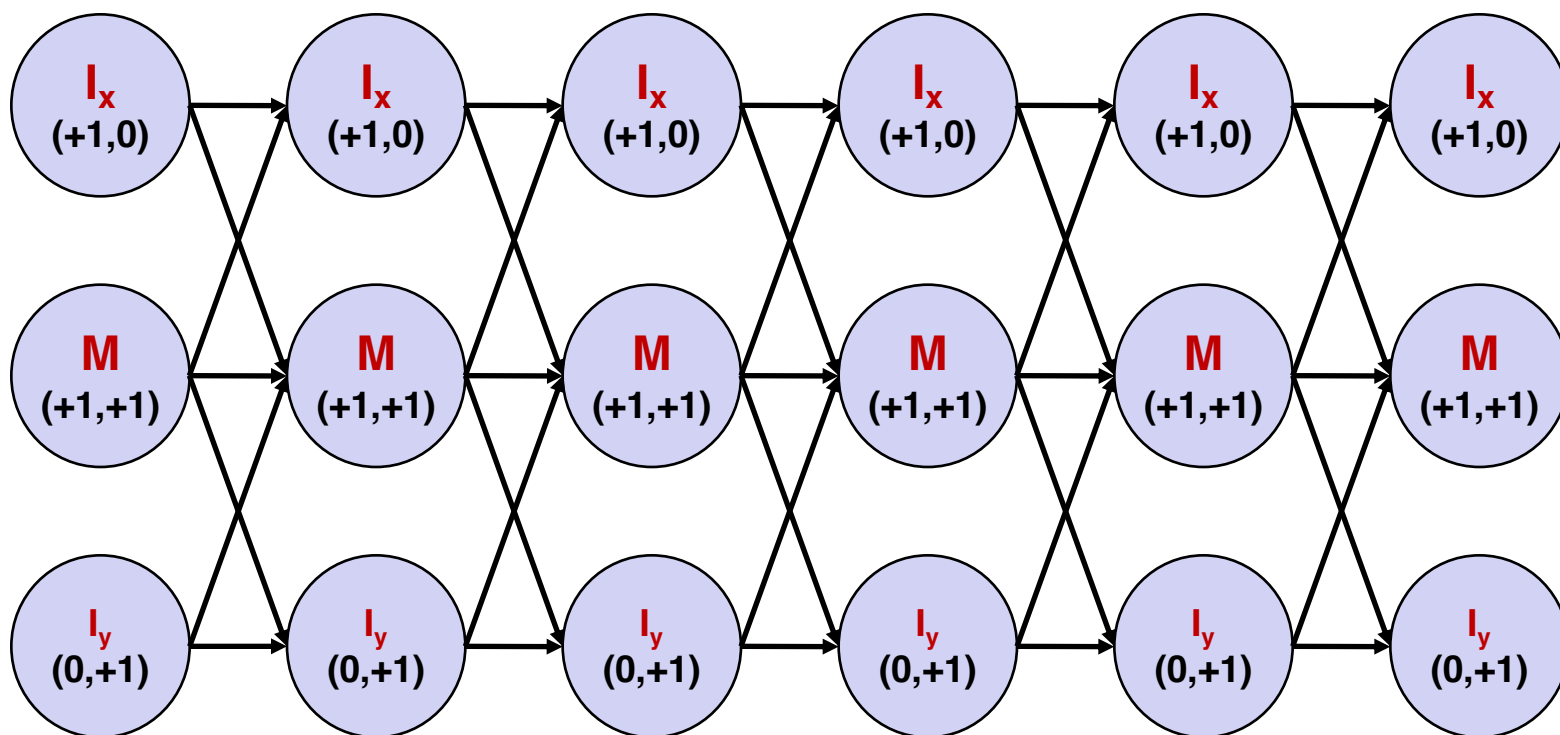
# State Transition Trellis



# State Transition Trellis

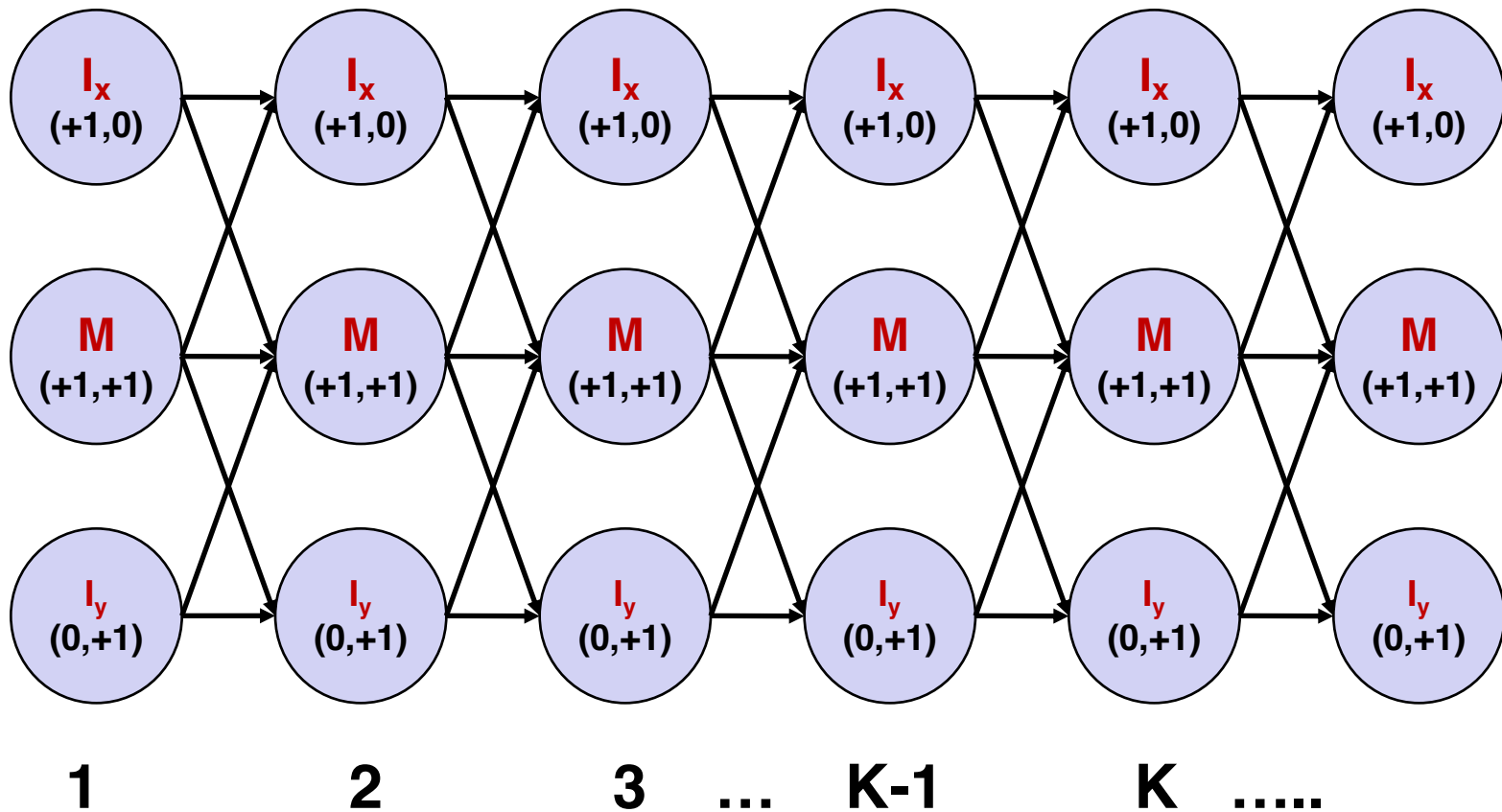
---

But how do we find the best alignment?



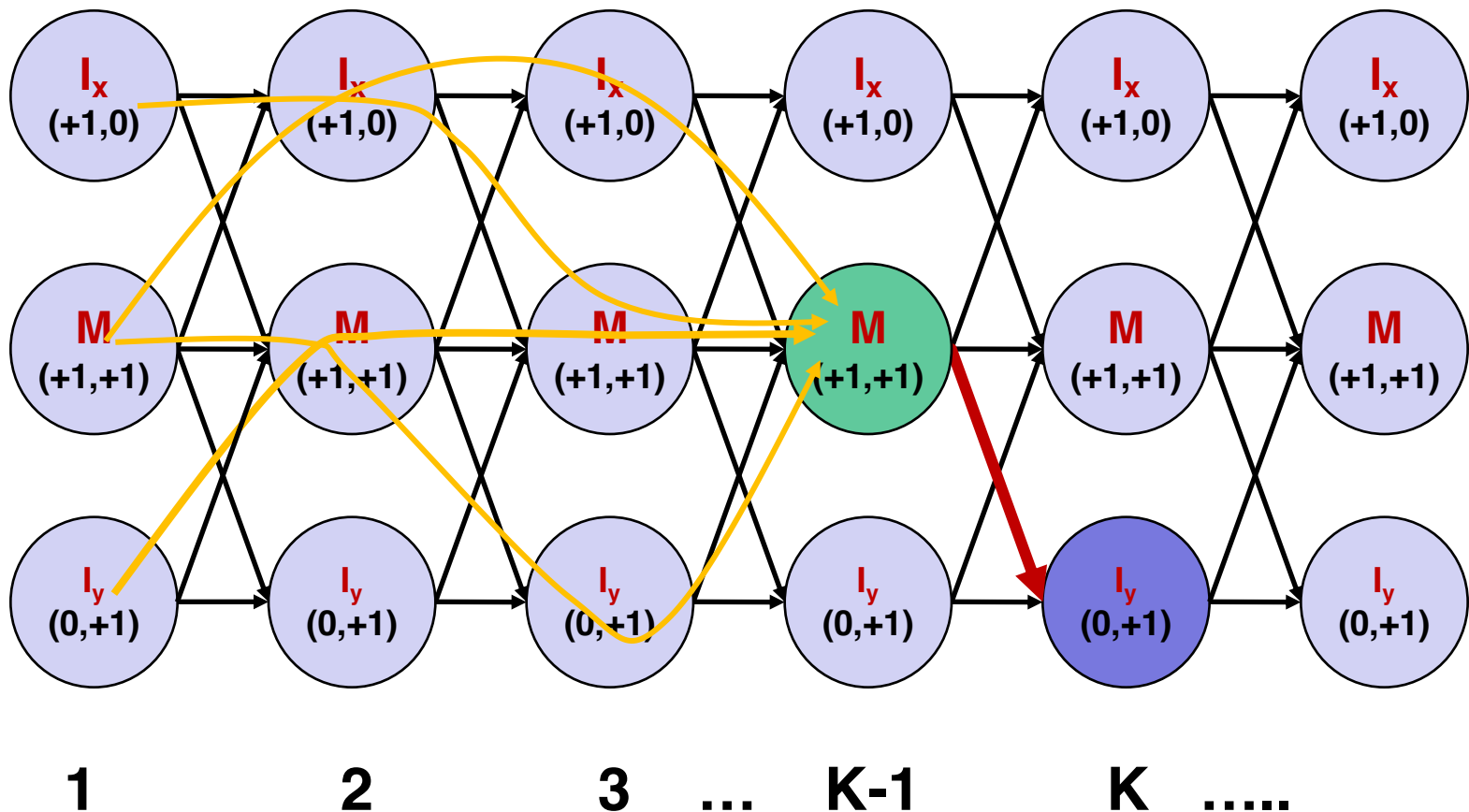
# State Transition Trellis

Define  $I_{x,k}$  as score of best alignment to *position*  $K$  ending in state  $I_x$   
 $I_{y,k}$  as score of best alignment to *position*  $K$  ending in state  $I_y$   
 $M_k$  as score of best alignment to *position*  $K$  ending in state  $M$



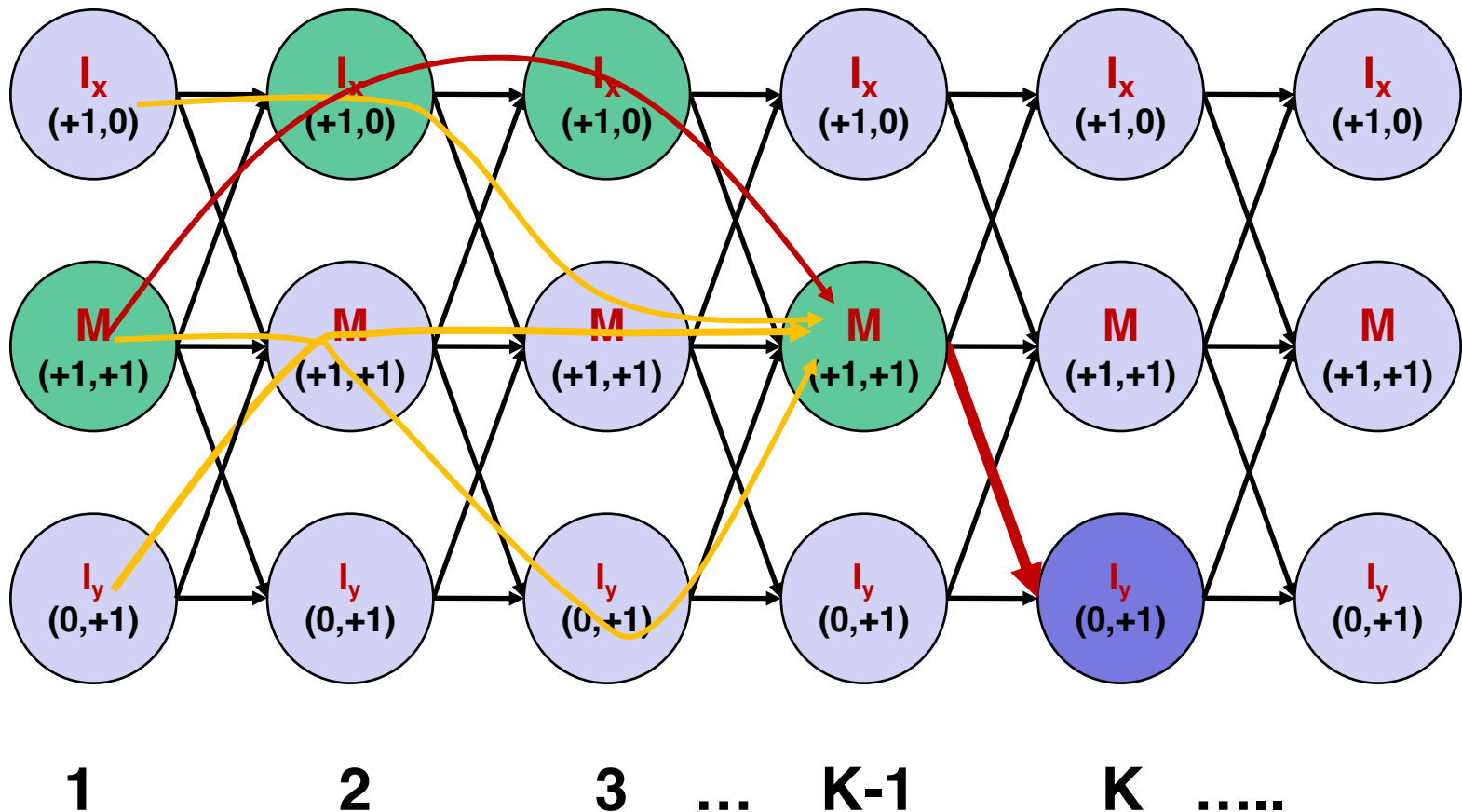
# Optimality Substructure Revisted

Define  $I_{x,k}$  as score of best alignment to *position*  $K$  ending in state  $I_x$   
 $I_{y,k}$  as score of best alignment to *position*  $K$  ending in state  $I_y$   
 $M_k$  as score of best alignment to *position*  $K$  ending in state  $M$



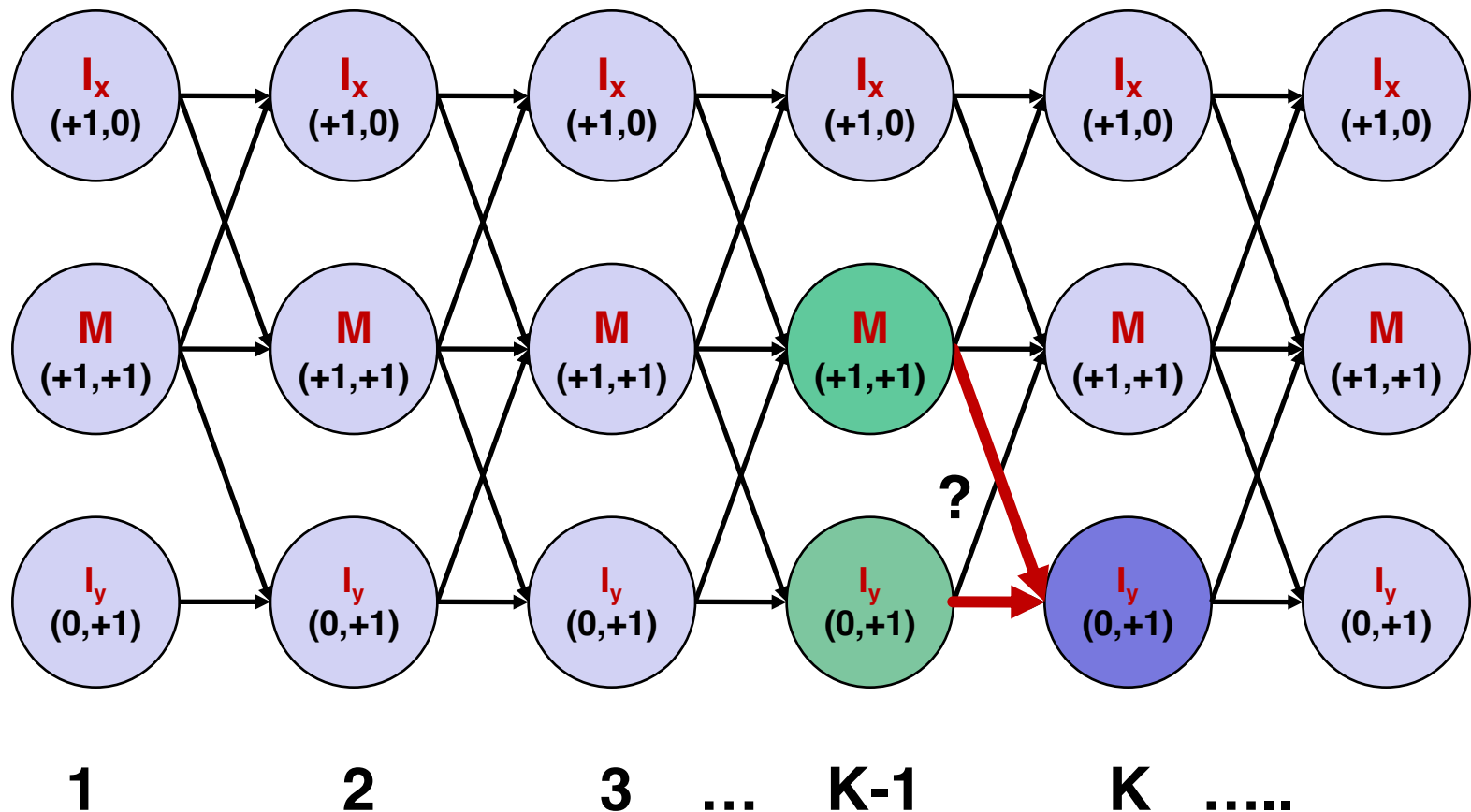
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Define  $I_{x,k}$  as score of best alignment to *position*  $K$  ending in state  $I_x$   
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# Optimality Substructure Revisted

Define  $I_{x,k}$  as score of best alignment to *position*  $K$  ending in state  $I_x$   
 $I_{y,k}$  as score of best alignment to *position*  $K$  ending in state  $I_y$   
 $M_k$  as score of best alignment to *position*  $K$  ending in state  $M$



# Finite State Machine

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## Trellis

$$M_k = \max \begin{cases} M_{k-1} + s(x, y) \\ I_{x,k-1} + s(x, y) \\ I_{y,k-1} + s(x, y) \end{cases}$$

$$I_{x,k} = \max \begin{cases} M_{k-1} - d \\ I_{x,k-1} - e \end{cases}$$

$$I_{y,k} = \max \begin{cases} M_{k-1} - d \\ I_{y,k-1} - e \end{cases}$$

## F Matrix

$$M(i, j) = \max \begin{cases} M(i-1, j-1) + s(x_i, y_j) \\ I_x(i-1, j) + s(x_i, y_j) \\ I_y(i, j-1) + s(x_i, y_j) \end{cases}$$

$$I_x(i, j) = \max \begin{cases} M(i-1, j) - d \\ I_x(i-1, j) - e \end{cases}$$

$$I_y(i, j) = \max \begin{cases} M(i, j-1) - d \\ I_y(i, j-1) - e \end{cases}$$

**Equivalent ways of looking at the same problem.**  
**Parameterized over different spaces: (i,j) vs (k)**

