Semi-global and local alignment and gap penalties

Maximization vs. Minimization

Edit distance:

$$\text{OPT}(i, j) = \min \begin{cases} \text{cost}(x_i, y_j) + \text{OPT}(i - 1, j - 1) & \textbf{match } \textbf{x}_\textbf{i}, \textbf{y}_\textbf{j} \\ \text{c}_\text{gap} + \text{OPT}(i - 1, j) & \textbf{x}_\textbf{i} \textbf{ is unmatched} \\ \text{c}_\text{gap} + \text{OPT}(i, j - 1) & \textbf{y}_\textbf{j} \textbf{ is unmatched} \end{cases}$$
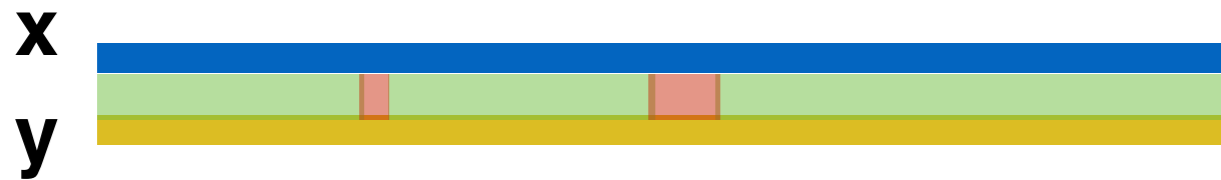
Sequence Similarity: replace the *min* with a *max* — find the highest-scoring alignment. Gap costs and bad matches usually get a negative "score".

$$\text{OPT}(i, j) = \max \begin{cases} \text{score}(x_i, y_j) + \text{OPT}(i - 1, j - 1) \\ \text{s}_\text{gap} + \text{OPT}(i - 1, j) \\ \text{s}_\text{gap} + \text{OPT}(i, j - 1) \end{cases}$$

gap penalty → gap score (probably negative)
match cost → match score

*

Alignment Categories

Global: Require an end-to-end alignment of **x**,**y**

x
y

Semi-global (glocal): Gaps at the beginning or end of **x** or **y** are free — useful when one string is significantly shorter than the other or for finding overlaps between strings
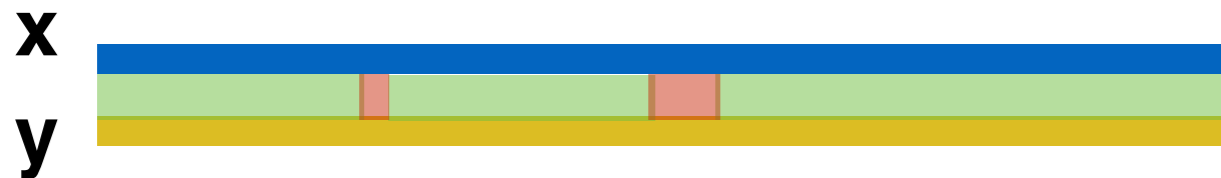
x
y

or

x
y

Local: Find the highest scoring alignment between **x'** a substring of **x** and **y'** a substring of **y** — useful for finding similar regions in strings that may not be globally similar
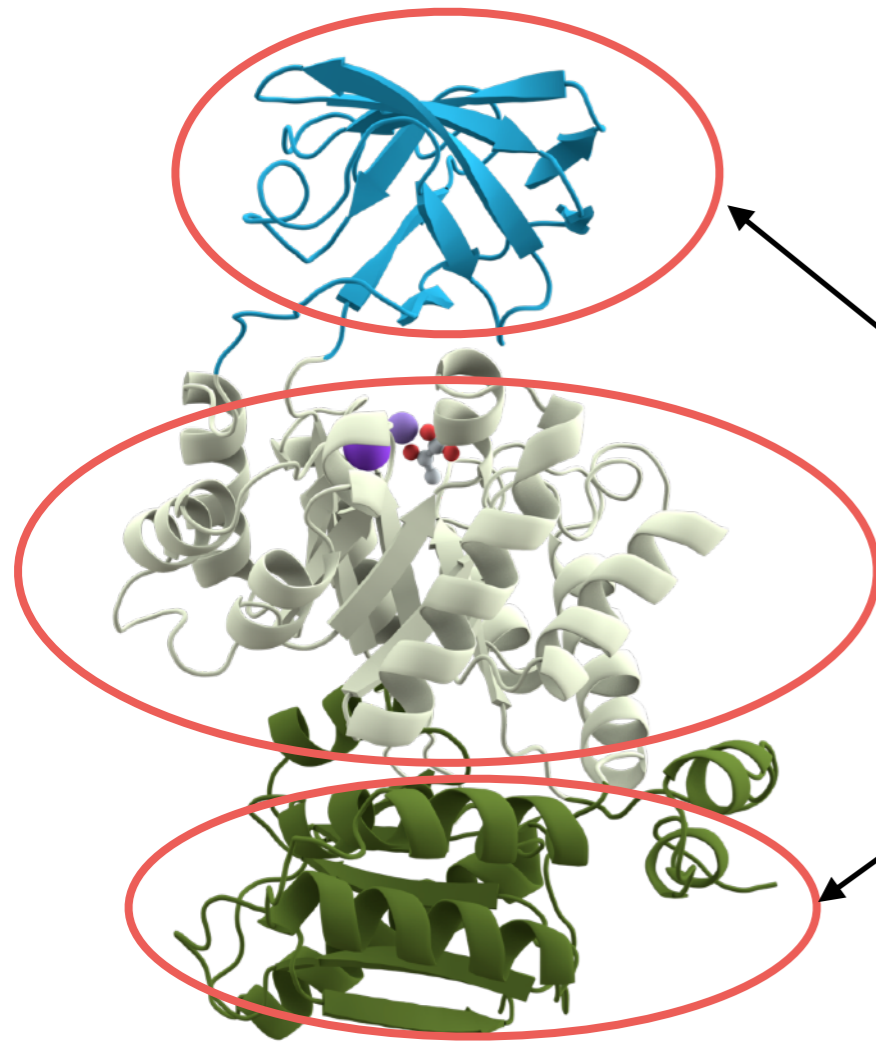
x
y

Alignment Categories Motivation

Global: **x** and **y** are similar proteins from closely-related species

x

y

Semi-global (glocal): **x** and **y** are sequencing reads we are trying to assemble. We want to find reads where the right end (suffix) of one matches the left end (prefix) of another.

x

y

or

x

y

Alignment Categories Motivation

It's possible and somewhat common for specific domains to be conserved, but not the entire protein sequence / structure.
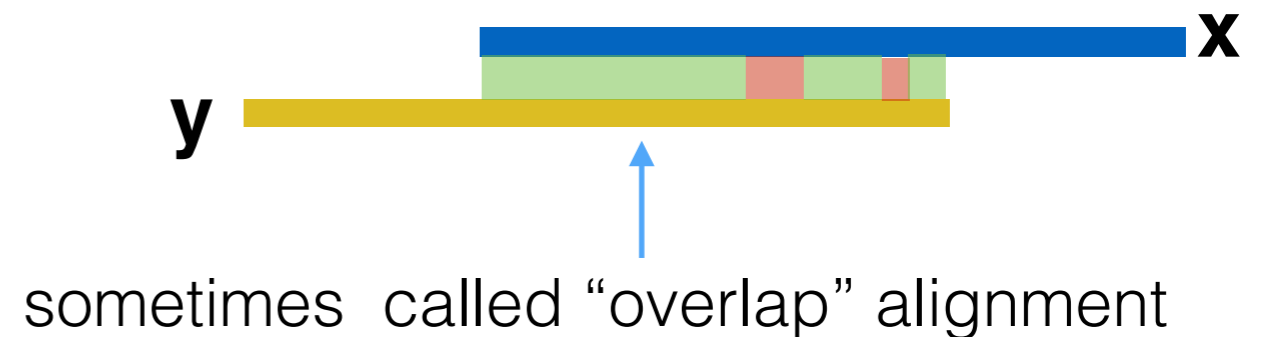
Local: **x** and **y** are similar proteins from potentially distantly related species. We don't expect the entire protein to be conserved, but certain "domains" should be.

Semi-global Alignment Example

Semi-global (glocal): Gaps at the beginning or end of **x** or **y** are free. Useful when one one string is significantly shorter than the other or we want to find an overlap between the suffix of one string and a prefix of the other

sometimes called "cost-free-ends" or "fitting" alignment

sometimes called "overlap" alignment

Motivation:

Useful for finding similarities that global alignments wouldn't. Also, can view "read mapping" as a variant of the semi-global alignment problem. Want to align entire read but it's a tiny fraction of the genome. *Note*: won't use semi-global alignment with the full genome for read mapping in practice.

Semi-global Alignment Example

Semi-global (glocal): Gaps at the beginning or end of **x** or **y** are free — one useful case is when one string is significantly shorter than the other

sometimes called "cost-free-ends" or "fitting" alignment

We'll discuss the "fitting" variant for in the next few slides for simplicity, but the same basic idea applies for the "overlap" variant as well.
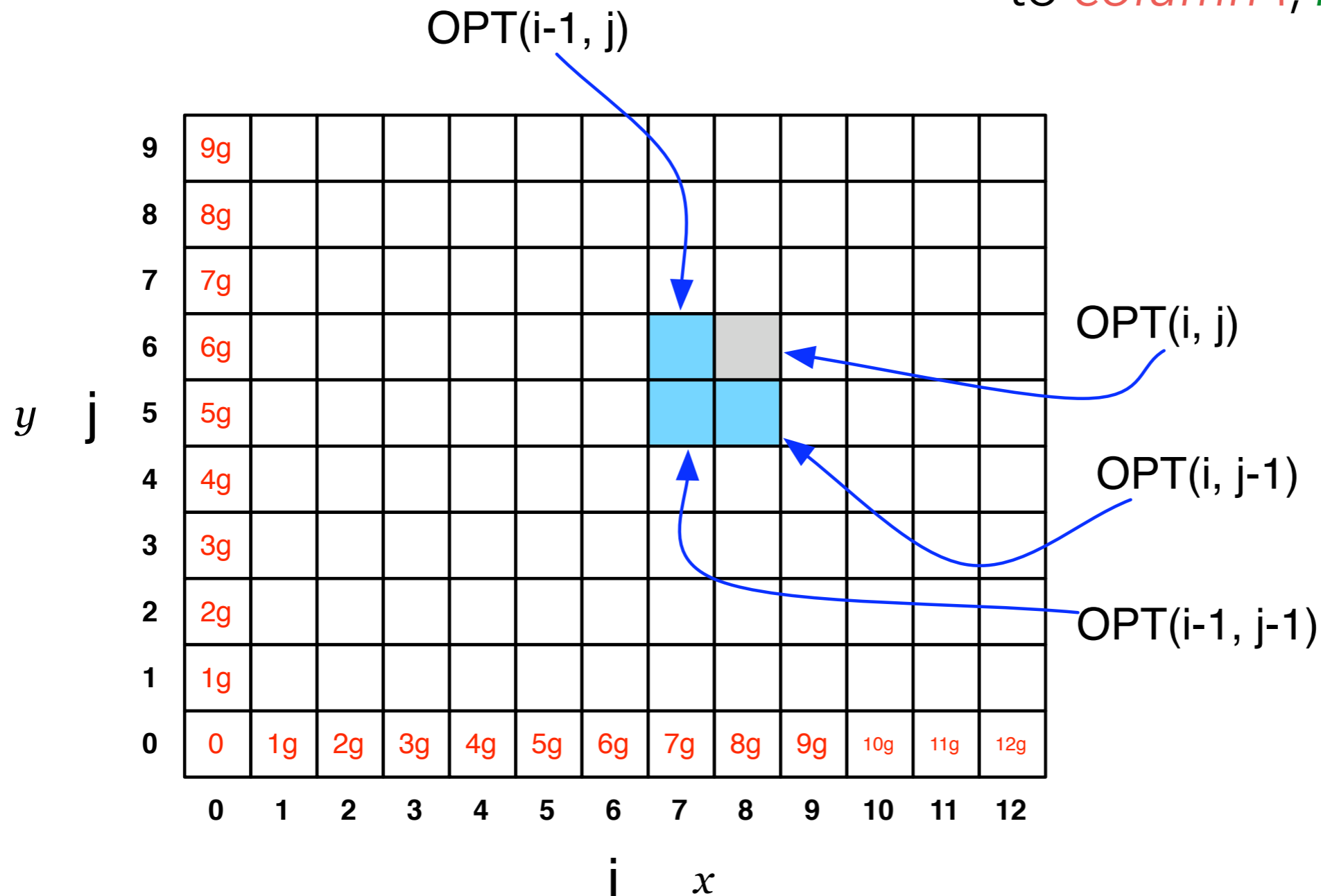
Recall: Global Alignment Matrix

OPT(i,j) contains the score for the best alignment between:

the first *i* characters of string *x* [**prefix** *i* of *x*]

the first *j* character of string *y* [**prefix** *j* of *y*]

NOTE: observe the non-standard notation here; OPT(**i**,**j**) is referring to *column i*, *row j* of the matrix.

OPT(i-1, j)

OPT(i, j)

OPT(i, j-1)

OPT(i-1, j-1)

How to do semi-global alignment?

y

$m \cdot s_{gap}$										
$3 \cdot s_{gap}$										
$2 \cdot s_{gap}$										
$1 \cdot s_{gap}$										
0	$1 \cdot s_{gap}$	$2 \cdot s_{gap}$	$3 \cdot s_{gap}$							$n \cdot s_{gap}$

x

Start with the original global alignment matrix

How to do semi-global alignment?

y

$m \cdot s_{gap}$												
$3 \cdot s_{gap}$												
$2 \cdot s_{gap}$												
$1 \cdot s_{gap}$												
0	0	0	0									0

x

change the base case — allow gaps *before* y

How to do semi-global alignment?

y

m·s$_{gap}$									O(n,m)
3·s$_{gap}$									
2·s$_{gap}$									
1·s$_{gap}$									
0	0	0	0						0

x

start traceback at $\max_{0<i\leq n}$ OPT(i,m) — this allows gaps after **y**; why?

Semi-global alignment example

and this gap after **y**

y

$m \cdot s_{gap}$

$O(n,m)$

$3 \cdot s_{gap}$

$2 \cdot s_{gap}$

$1 \cdot s_{gap}$

0   0   0   0   0

x

We allow this gap before **y**

Semi-global Alignment

What is the same and different between the "global" and semi-global ("fitting") alignment problems?

*assuming $|y| < |x|$ and we are "fitting" y into x

Global	Semi-global ("fitting")

$$\text{OPT}(i,j) = \max \begin{cases} \text{score}(x_i, y_j) + \text{OPT}(i-1, j-1) \\ \text{s}_\text{gap} + \text{OPT}(i-1, j) \\ \text{s}_\text{gap} + \text{OPT}(i, j-1) \end{cases}$$

$$\text{OPT}(i,j) = \max \begin{cases} \text{score}(x_i, y_j) + \text{OPT}(i-1, j-1) \\ \text{s}_\text{gap} + \text{OPT}(i-1, j) \\ \text{s}_\text{gap} + \text{OPT}(i, j-1) \end{cases}$$

Base case: OPT(i,0) = i x s_gap

Base case: OPT(i,0) = 0

Traceback starts at OPT(n,m)

Traceback starts at **max** OPT(j,m)
$_{0<j\leq n}$

Semi-global Alignment

The recurrence remains the *same*, we only change the base case of the recurrence and the origin of the backtrack

1) Ignore gaps before x ⟶ change base case;
   $$OPT(0,j) = 0$$

2) Ignore gaps after x ⟶ change traceback;
   start from $\max_{0<j\leq m} OPT(n,j)$

3) Ignore gaps before y ⟶ change base case;
   $$OPT(i,0) = 0$$

4) Ignore gaps after y ⟶ change traceback;
   start from $\max_{0<i\leq n} OPT(i,m)$

Semi-global Alignment

1) Ignore gaps before x

2) Ignore gaps after x

3) Ignore gaps before y

4) Ignore gaps after y
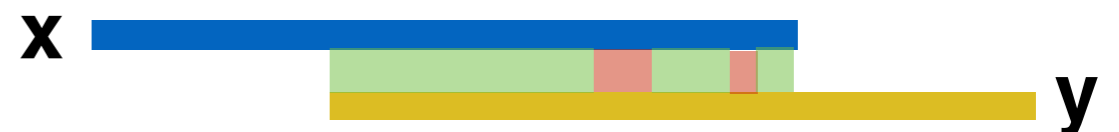
Types of semi-global alignments

use mods 3&4

use mods 1&4

use mods 1&2
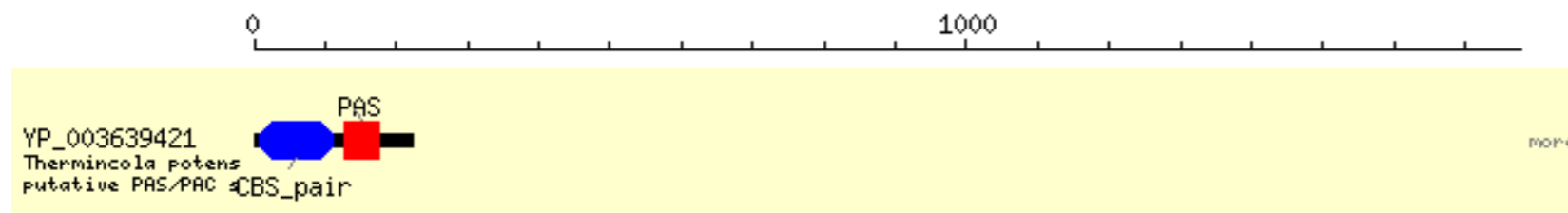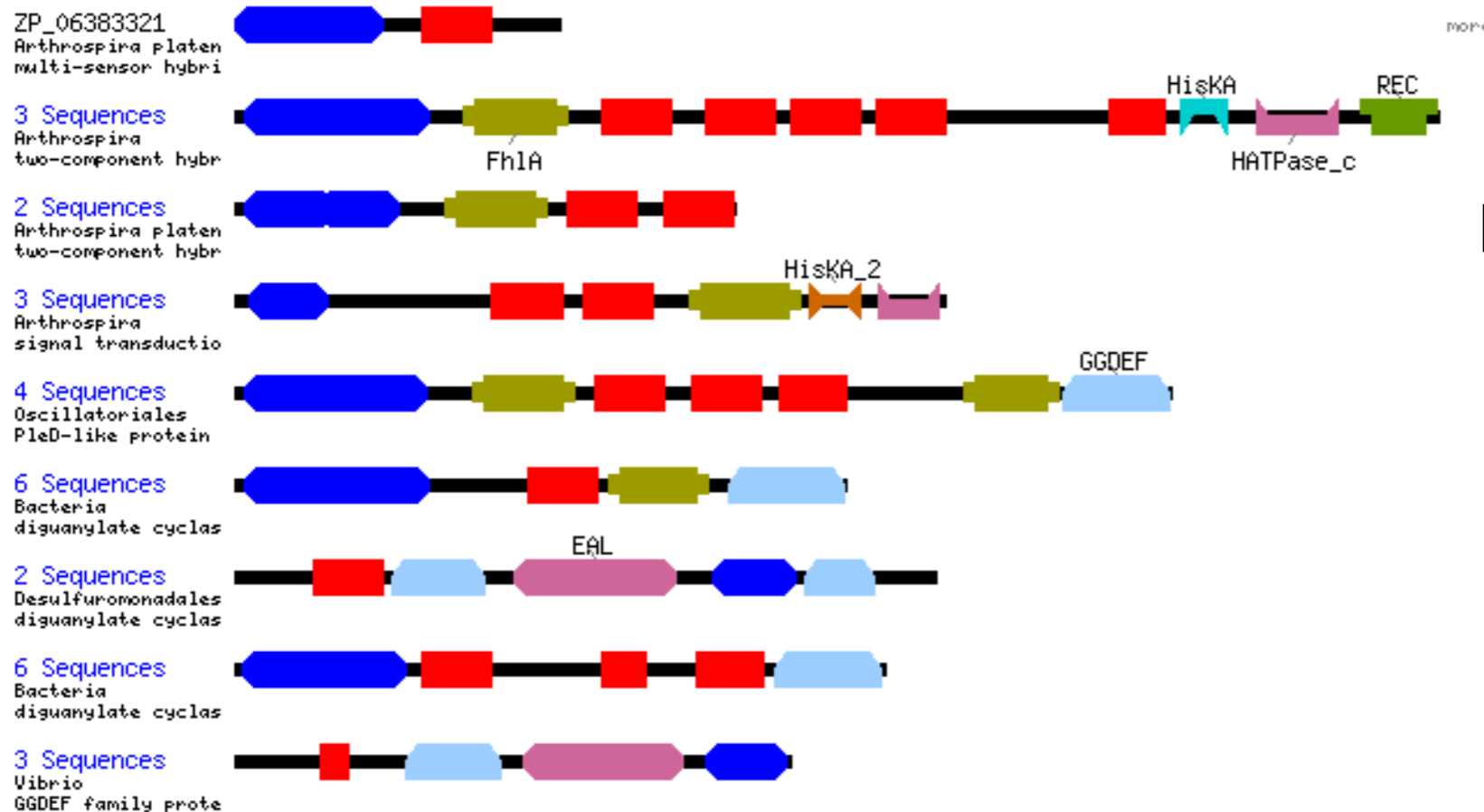
use mods 2&3

Local Alignment

a

b

Local alignment between a and b: Best alignment between a subsequence of a and **a** subsequence of **b**.

Motivation:
Many genes are composed of *domains*, which are subsequences that perform a particular function.

*

Local Alignment

Best alignment between
a suffix of x[1..5] and a
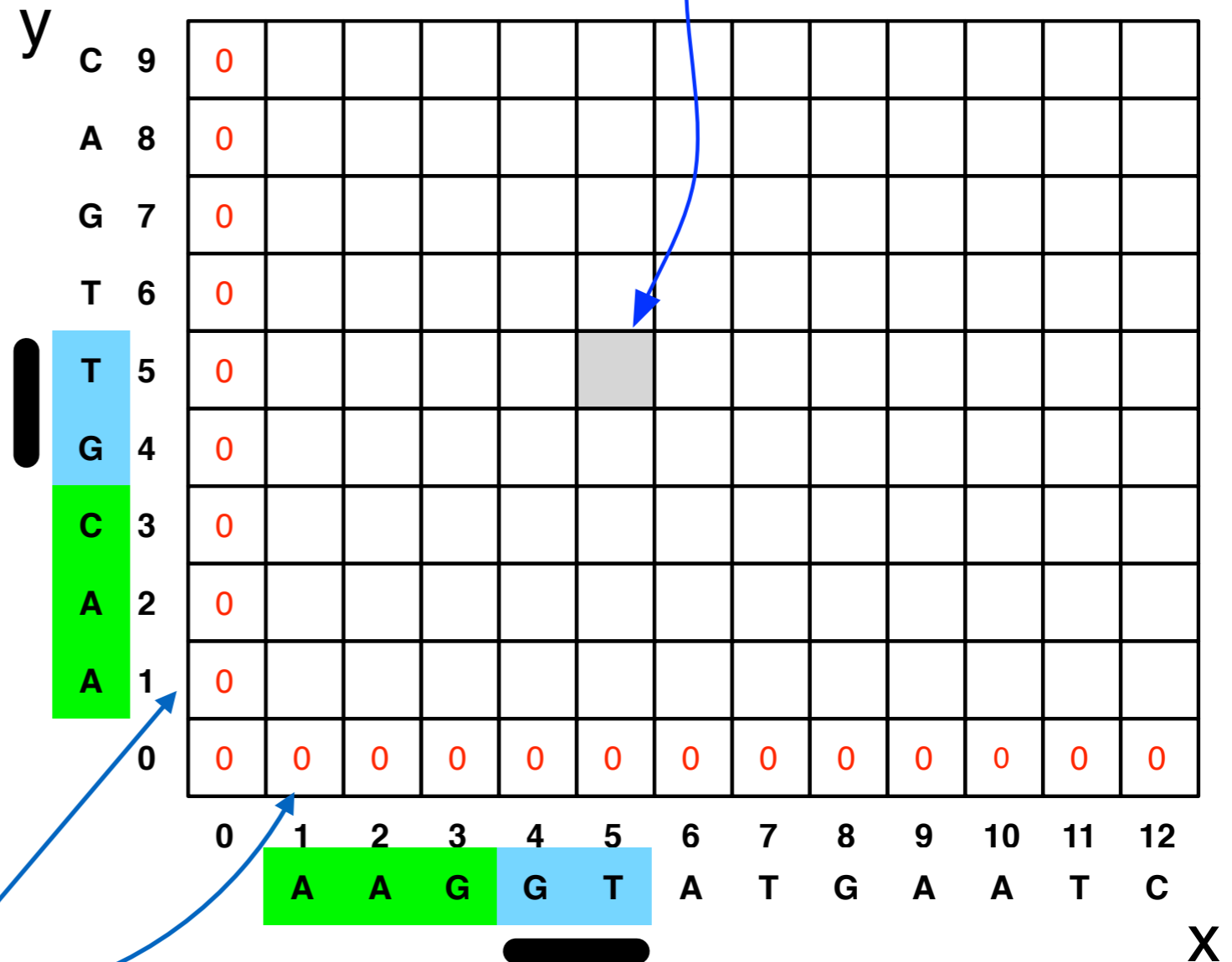suffix of y[1..5]

New meaning of entry of
matrix entry:

OPT(i, j) = best score
between:

   some suffix of $x[1...i]$

   some suffix of $y[1...j]$

Same base-case
trick we used in semi-global alignment

y

		0											
C	9	0											
A	8	0											
G	7	0											
T	6	0											
T	5	0											
G	4	0											
C	3	0											
A	2	0											
A	1	0											
	0	0	0	0	0	0	0	0	0	0	0	0	0

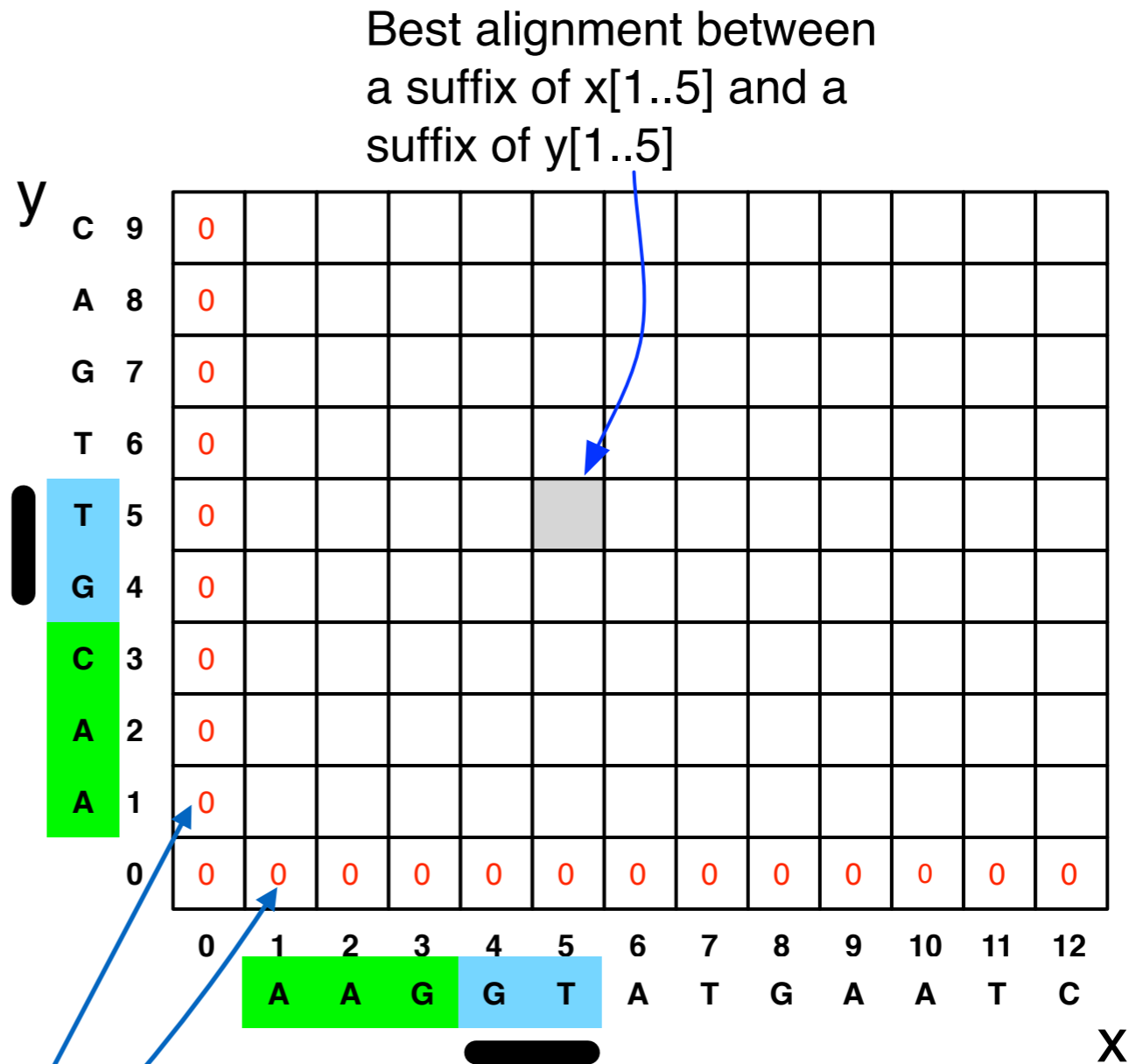|	0	1	2	3	4	5	6	7	8	9	10	11	12
		A	A	G	G	T	A	T	G	A	A	T	C

x

*

Local Alignment

New meaning of entry of matrix entry:

OPT(i, j) = best score between:
  some suffix of $x[1...i]$
  some suffix of $y[1...j]$

What else do we need to change to allow local alignments?

Hint: The empty alignment is always a valid local alignment!

Best alignment between a suffix of x[1..5] and a suffix of y[1..5]

Same base-case
trick we used in semi-global alignment

*

How do we fill in the local alignment matrix?

$$\mathrm{OPT}(i,j) = \max \begin{cases} \mathrm{score}(x_i, y_j) + \mathrm{OPT}(i-1, j-1) & \text{(1)} \\ \mathrm{s_{gap}} + \mathrm{OPT}(i-1, j) & \text{(2)} \\ \mathrm{s_{gap}} + \mathrm{OPT}(i, j-1) & \text{(3)} \\ 0 \end{cases}$$

(1), (2), and (3): same cases as before: match x and y, gap in y, gap in x

New case: 0 allows you to say the best alignment between a suffix of *x* and a suffix of *y* is the empty alignment.

Lets us "start over"

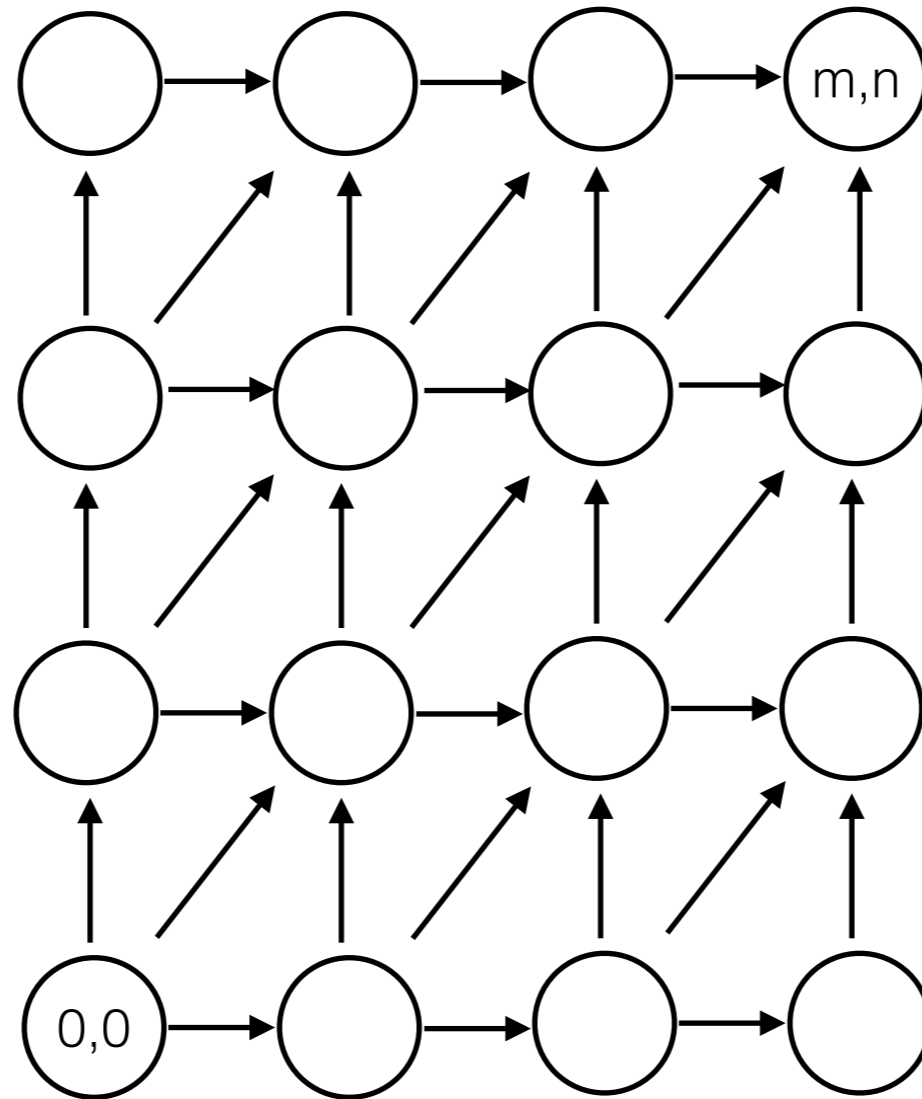Best alignment between a suffix of x[1..5] and a suffix of y[1..5]
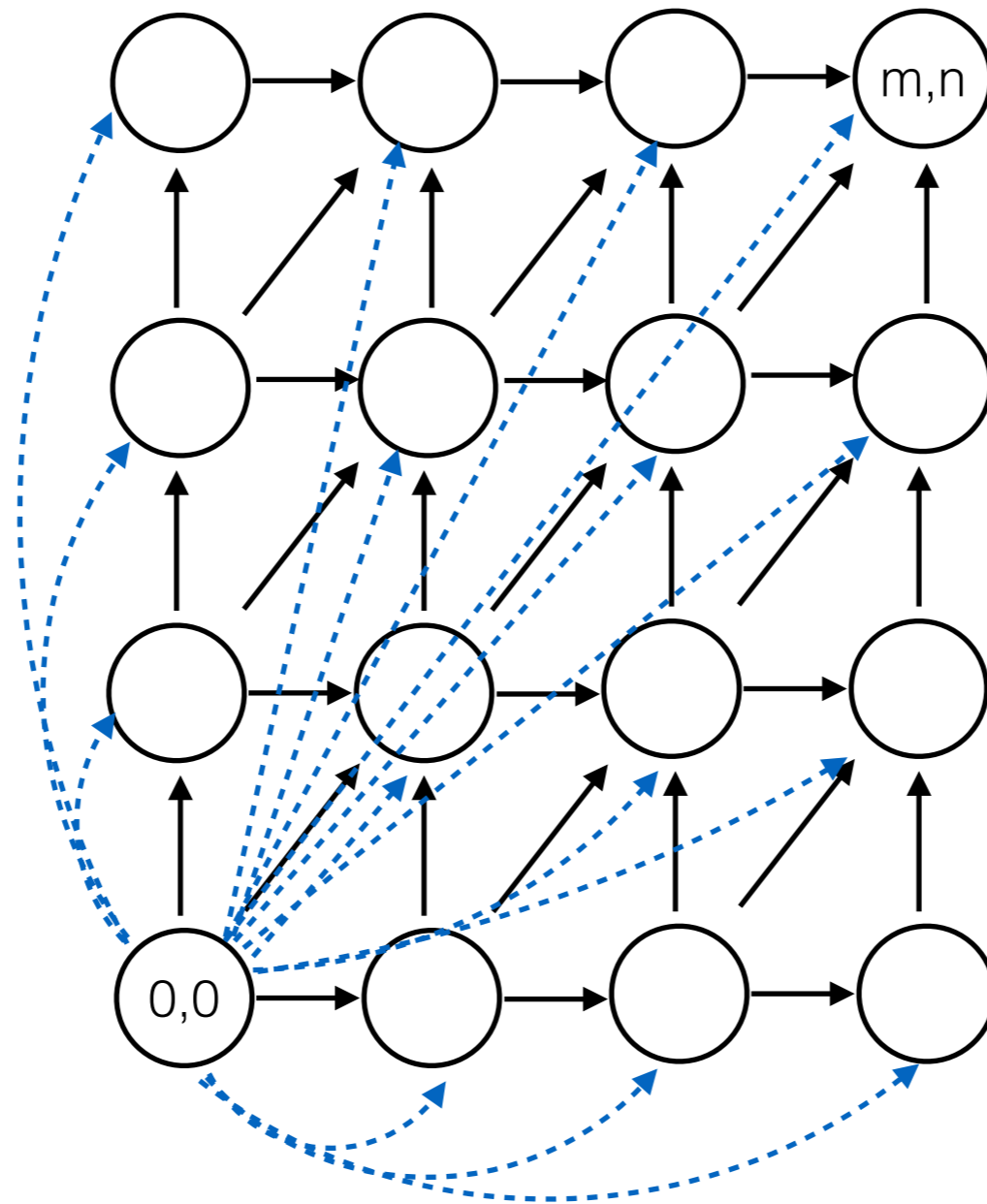
Local Alignment

- Initialize first row and first column to be 0.

- The score of the best local alignment is the largest value in the entire array.

- To find the actual local alignment:
  - start at an entry with the maximum score
  - traceback as usual
  - stop when we reach an entry with a score of 0

*

Local Alignment in the DAG framework
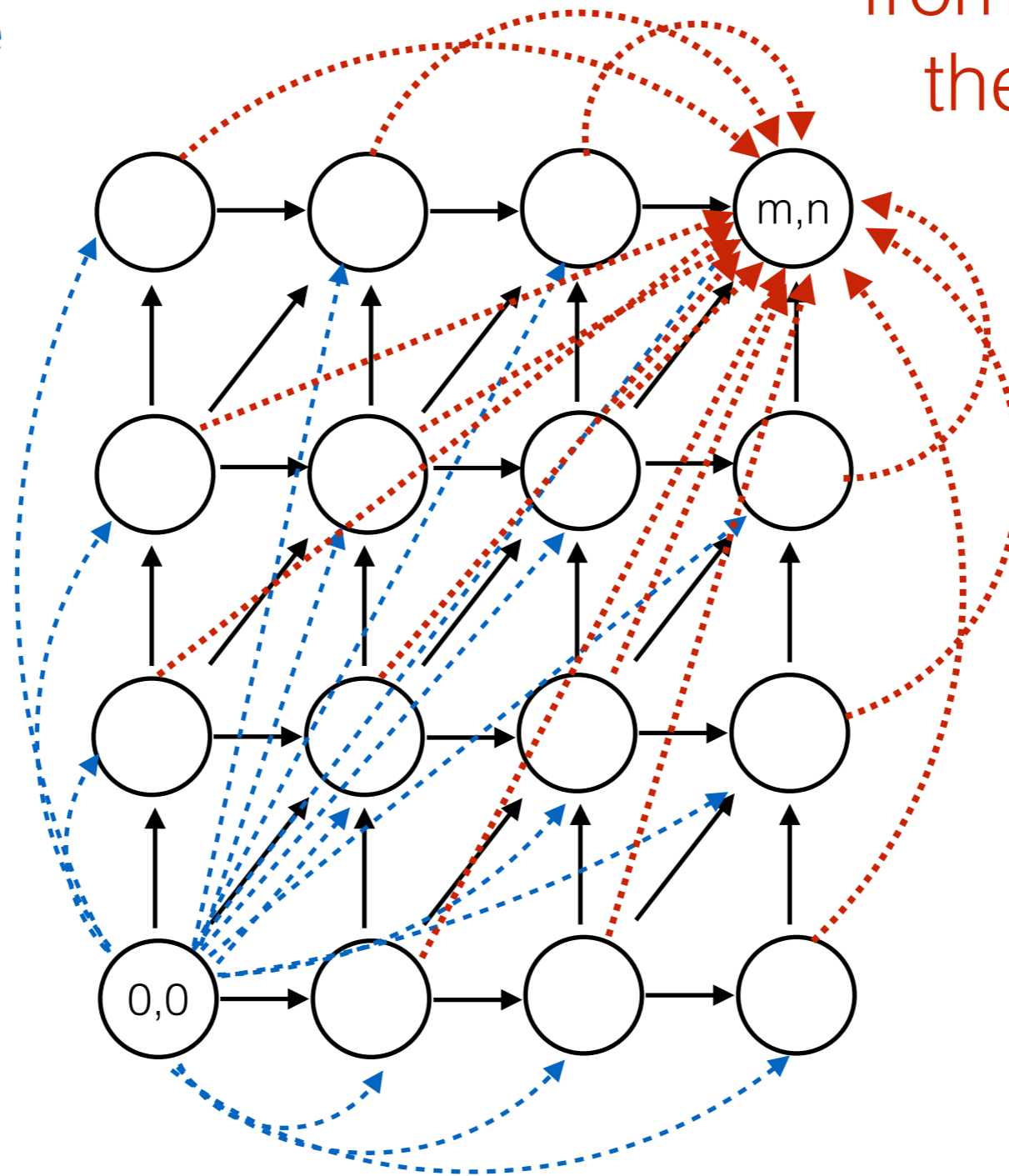
Local Alignment in the DAG framework

Add 0 score edge from the source to every node

Local Alignment in the DAG framework

Add 0 score edge from the source to every vertex

Add 0 score edge from every vertex to the target vertex

Local Alignment Example #1

local align("AGCGTAG", "CTCGTC")

	*	A	G	C	G	T	A	G
*	0	0	0	0	0	0	0	0
C	0	0	0	10	3	0	0	0
T	0	0	0	3	5	13	6	0
C	0	0	0	10	3	6	8	1
G	0	0	10	3	20	13	6	18
T	0	0	3	5	13	**30**	23	16
C	0	0	0	13	6	23	25	18

Score(match) = 10
Score(mismatch) = -5
Score(gap) = -7

Note: this table written top-to-bottom instead of bottom-to-top

*

Local Alignment Example #2

`local align("bestoftimes", "soften")`

	*	b	e	s	t	o	f	t	i	m	e	s
*	0	0	0	0	0	0	0	0	0	0	0	0
s	0	0	0	10	3	0	0	0	0	0	0	10
o	0	0	0	3	5	13	6	0	0	0	0	3
f	0	0	0	0	0	6	23	16	9	2	0	0
t	0	0	0	0	10	3	16	33	26	19	12	5
e	0	0	10	3	3	5	9	26	28	21	29	22
n	0	0	3	5	0	0	2	19	21	23	22	24

Score(match) = 10
Score(mismatch) = –5
Score(gap) = –7

Note: this table written top-to-bottom instead of bottom-to-top

*

More Local Alignment Examples

local align("catdogfish", "dog")

	*	c	a	t	d	o	g	f	i	s	h
*	0	0	0	0	0	0	0	0	0	0	0
d	0	0	0	0	10	3	0	0	0	0	0
o	0	0	0	0	3	20	13	6	0	0	0
g	0	0	0	0	0	13	**30**	23	16	9	2

local align("mississippi", "issp")

	*	m	i	s	s	i	s	s	i	p	p	i
*	0	0	0	0	0	0	0	0	0	0	0	0
i	0	0	10	3	0	10	3	0	10	3	0	10
s	0	0	3	20	13	6	20	13	6	5	0	3
s	0	0	0	13	30	23	16	30	23	16	9	2
p	0	0	0	6	23	25	18	23	25	**33**	26	19

local align("aaaa", "aa")

	*	a	a	a	a
*	0	0	0	0	0
a	0	10	10	10	10
a	0	10	**20**	**20**	**20**

*

Local / Global Recap

- Alignment score sometimes called the "edit distance" between two strings.

- Edit distance is sometimes called Levenshtein distance.

- Algorithm for local alignment is sometimes called "Smith-Waterman"

- Algorithm for global alignment is sometimes called "Needleman-Wunsch"

- Same basic algorithm, however.

- Underlies BLAST

*

General Gap Penalties

```
AAAGAATTCA                    AAAGAATTCA
A-A-A-T-CA        vs.         AAA----TCA
```

These have the same score, but the second one is often more plausible.

A single insertion of "GAAT" into the first string could change it into the second — Biologically, this is much more likely as **x** could be transformed into **y** in "one fell swoop".

- Currently, the score of a run of k gaps is $s_{gap} \times k$

- It might be more realistic to support general gap penalty, so that the score of a run of k gaps is $|\textbf{gscore}(k)| < |(s_{gap} \times k)|$.

- Then, the optimization will prefer to group gaps together.

*

General Gap Penalties — The Problem

```
AAAGAATTCA              AAAGAATTCA
A-A-A-T-CA    vs.       AAA----TCA
```

Previous DP no longer works with general gap penalties.

Why?

*

General Gap Penalties — The Problem

```
AAAGAATTCA              AAAGAATTCA
A-A-A-T-CA      vs.     AAA----TCA
```

The score of the *last character* depends on *details* of the previous alignment:

```
AAAGAAC                 AAAGAATC
AAA----         vs.     AAA----
```

We need to "know" how long a final run of gaps is in order to give a score to the last subproblem.

General Gap Penalties — The Problem

The score of the *last character* depends on *details* of the previous alignment:

Knowing the optimal alignment at the substring ending here.

```
AAAGAAC                AAAGAATC
AAA----     vs.        AAA-----
```

Doesn't let us simply build the optimal alignment ending here.

*

Three Matrices

We now keep 3 different matrices:

M(i,j) = score of best alignment of x[1..i] and y[1..j] ending with a character-character **match or mismatch**.

X(i,j) = score of best alignment of x[1..i] and y[1..j] ending with a **gap in X.**

Y(i,j) = score of best alignment of x[1..i] and y[1..j] ending with a **gap in Y.**

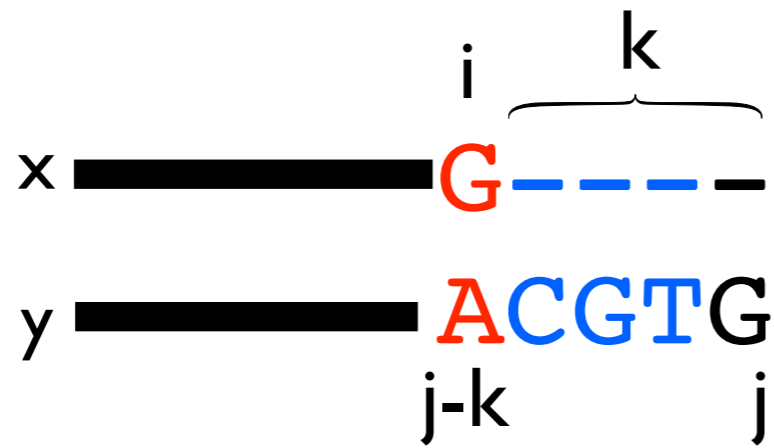$$\mathrm{M}(i,j) = \mathrm{score}(x_i, y_j) + \max \begin{cases} \mathrm{M}(i-1, j-1) \\ \mathrm{X}(i-1, j-1) \\ \mathrm{Y}(i-1, j-1) \end{cases}$$

$$\mathrm{X}(i,j) = \max \begin{cases} \mathrm{M}(i, j-k) + \mathrm{gscore}(k) & \text{for } 1 \leq k \leq j \\ \mathrm{Y}(i, j-k) + \mathrm{gscore}(k) & \text{for } 1 \leq k \leq j \end{cases}$$

$$\mathrm{Y}(i,j) = \max \begin{cases} \mathrm{M}(i-k, j) + \mathrm{gscore}(k) & \text{for } 1 \leq k \leq i \\ \mathrm{X}(i-k, j) + \mathrm{gscore}(k) & \text{for } 1 \leq k \leq i \end{cases}$$

*

The M Matrix

We now keep 3 different matrices:

M(i,j) = score of best alignment of x[1..i] and y[1..j] ending with a character-character **match or mismatch**.

X(i,j) = score of best alignment of x[1..i] and y[1..j] ending with a **gap in X**.

Y(i,j) = score of best alignment of x[1..i] and y[1..j] ending with a **gap in Y**.

By definition, alignment
ends in a match/mismatch.

$$\mathrm{M}(i,j) = \mathrm{score}(x_i, y_j) + \max \begin{cases} \mathrm{M}(i-1, j-1) \\ \mathrm{X}(i-1, j-1) \\ \mathrm{Y}(i-1, j-1) \end{cases}$$

Any kind of alignment is allowed
before the match/mismatch.

A

G

*

The X (and Y) matrices

k decides how long to make the gap.

We have to make the whole gap at once in order to know how to score it.

$$X(i,j) = \max \begin{cases} M(i, j-k) + \text{gscore}(k) & \text{for } 1 \leq k \leq j \\ Y(i, j-k) + \text{gscore}(k) & \text{for } 1 \leq k \leq j \end{cases}$$

This case is automatically handled.

*

Running Time for Gap Penalties

$$M(i,j) = \text{score}(x_i, y_j) + \max \begin{cases} M(i-1, j-1) \\ X(i-1, j-1) \\ Y(i-1, j-1) \end{cases}$$

$$X(i,j) = \max \begin{cases} M(i, j-k) + \text{gscore}(k) & \text{for } 1 \leq k \leq j \\ Y(i, j-k) + \text{gscore}(k) & \text{for } 1 \leq k \leq j \end{cases}$$

$$Y(i,j) = \max \begin{cases} M(i-k, j) + \text{gscore}(k) & \text{for } 1 \leq k \leq i \\ X(i-k, j) + \text{gscore}(k) & \text{for } 1 \leq k \leq i \end{cases}$$

Final score is max {M(n,m), X(n,m), Y(n,m)}.

How do you do the traceback?

Runtime:

- Assume |X| = |Y| = n for simplicity: $3n^2$ subproblems

- $2n^2$ subproblems take $O(n)$ time to solve (**because we have to try all k**)

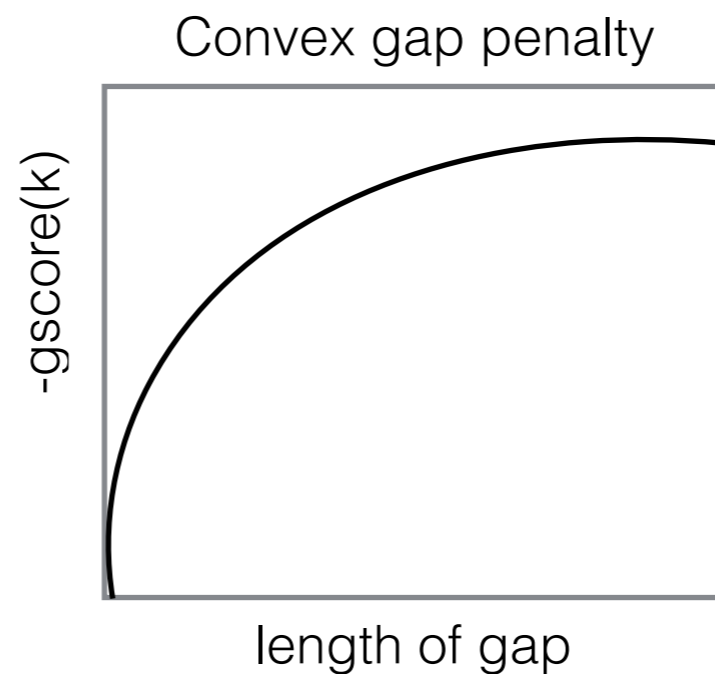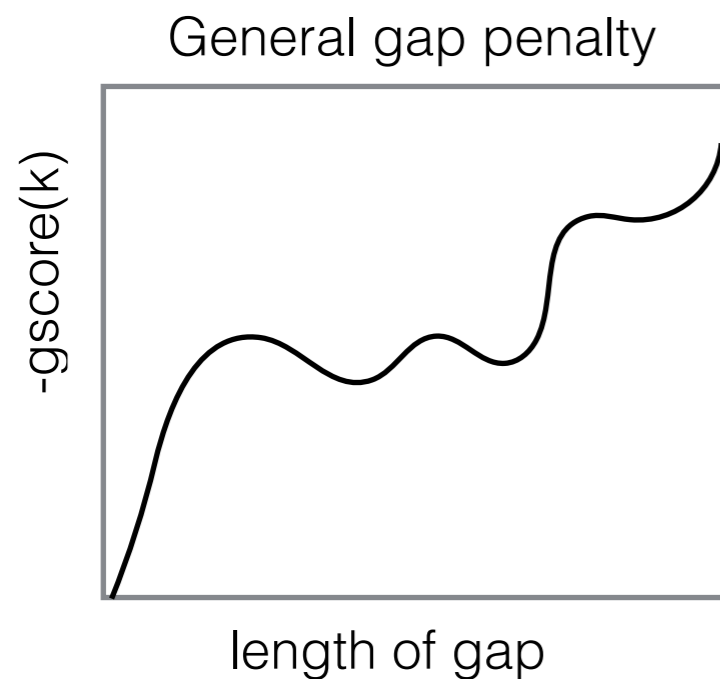$\Rightarrow$ $O(n^3)$ total time

*

Affine Gap Penalties

- $O(n^3)$ for general gap penalties is usually too slow...

- We can still encourage spaces to group together using a special case of general penalties called *affine gap penalties*:

  g_{start} = the cost of starting a gap

  g_{extend} = the cost of extending a gap by one more space
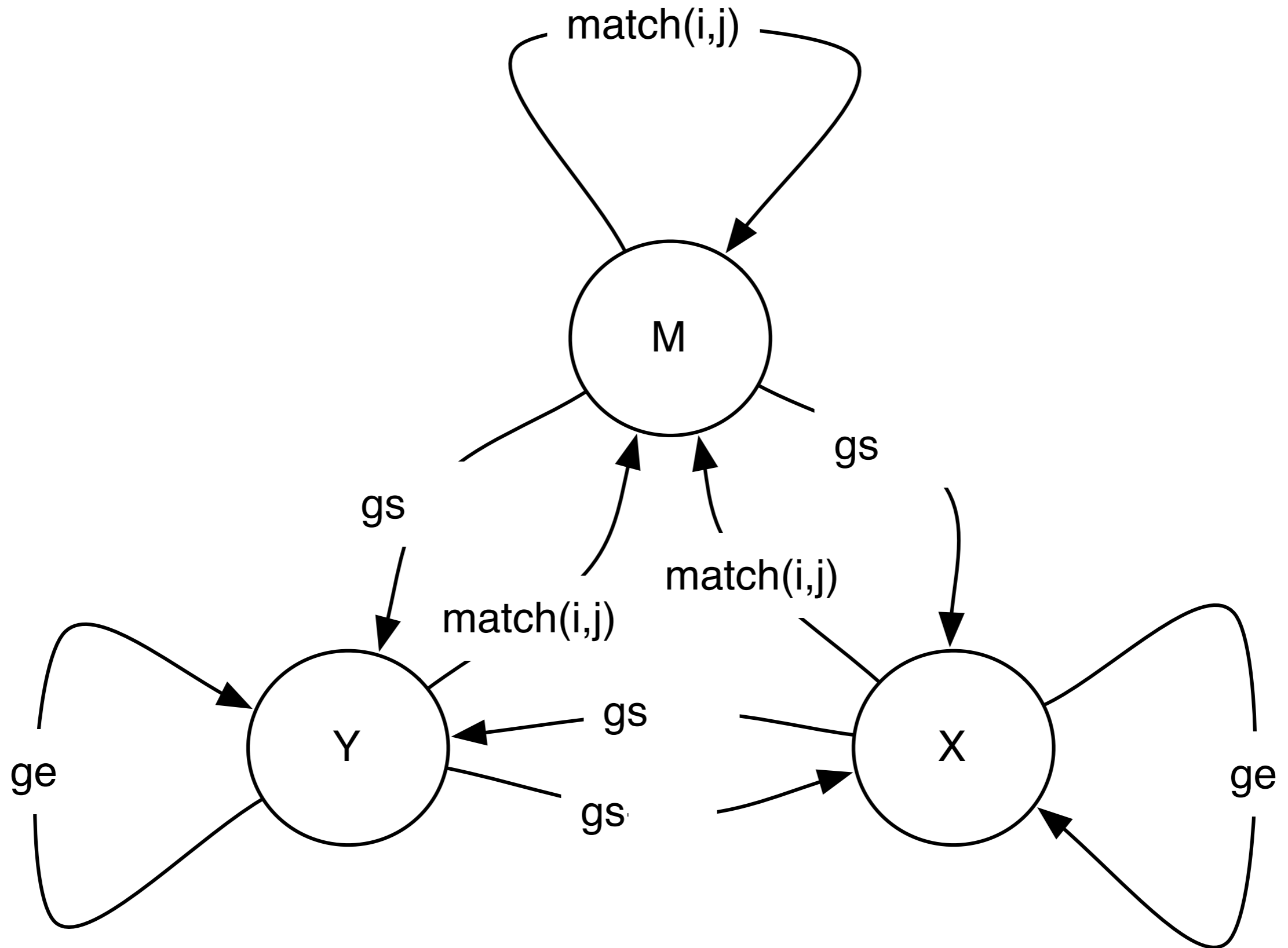
  $gscore(k) = g_{start} + (k-1) \times g_{extend}$

less restrictive $\Rightarrow$ more restrictive

General gap penalty

Convex gap penalty

Affine gap penalty

$(k-1)^*g_{extend}$

g_{start}

-gscore(k)

length of gap

Benefit of Affine Gap Penalties

- Same idea of using 3 matrices, but now we *don't need to search over all gap lengths*, we just have to know whether we are **starting a new gap** or **not**.

*

Affine Gap as Finite State Machine

Affine Gap Penalties

$$M(i,j) = \text{score}(x_i, y_i) + \max \begin{cases} M(i-1, j-1) \\ X(i-1, j-1) \\ Y(i-1, j-1) \end{cases}$$

(mis)match
between
x and y

If previous alignment ends in (mis)match, this is a new gap

$$X(i,j) = \max \begin{cases} g_{\text{start}} + M(i, j-1) \\ g_{\text{extend}} + X(i, j-1) \\ g_{\text{start}} + Y(i, j-1) \end{cases}$$

gap in x
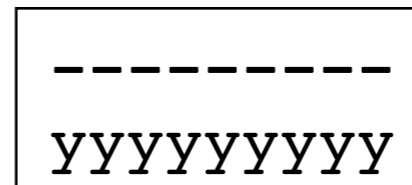
If we're using the X matrix, then we're extending a gap.

$$Y(i,j) = \max \begin{cases} g_{\text{start}} + M(i-1, j) \\ g_{\text{start}} + X(i-1, j) \\ g_{\text{extend}} + Y(i-1, j) \end{cases}$$
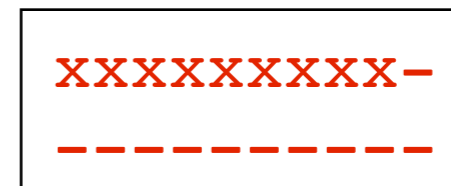
gap in y

If we're using the Y matrix, then we're starting a new gap in this string.

*

Affine Base Cases (Global)

- $M(0, i) =$ "score of best alignment between 0 characters of x and i characters of y that ends in a match" $= -\infty$ because no such alignment can exist.

- $X(0, i) =$ "score of best alignment between 0 characters of x and i characters of y that ends in a gap in x" $=$ gap_start $+$ (i-1) × gap_extend because this alignment looks like:

```
----------
yyyyyyyyyy
```

- $X(i, 0) =$ "score of best alignment between i characters of x and 0 characters of y that ends in a gap in X" $= -\infty$

```
xxxxxxxxx-          ← not allowed
----------
```

- $M(i, 0) = M(0, i)$ and $Y(0, i)$ and $Y(i, 0)$ are computed using the same logic as $X(i, 0)$ and $X(0, i)$

*

Affine Gap Runtime

- 3mn subproblems

- Each one takes **constant** time

- Total runtime O(mn):

  - back to the run time of the basic running time.

Traceback

- Arrows now can point **between** matrices.

- The possible arrows are given, as usual, by the recurrence.

  - E.g. What arrows are possible leaving a cell in the M matrix?
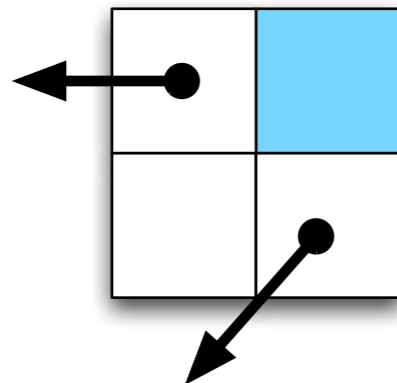
*

Why do you "need" 3 functions?

- Alternative WRONG algorithm:

```
M(i,j) = max(
    M(i-1, j-1) + cost(xᵢ, yⱼ),
    M(i-1, j) +(gstart if Arrow(i-1, j) != ←  , else gextend),
    M(j, i-1) + (gstart if Arrow(i, j-1) != ↓ , else gextend)
)
```

WRONG Intuition: we only need to know whether we are starting a gap or extending a gap.

The arrows coming out of each subproblem tell us how the best alignment ends, so we can use them to decide if we are starting a new gap.

The best alignment up to this cell ends in a gap.

The best alignment up to this cell ends in a match.

PROBLEM: The best alignment for strings x[1..i] and y[1..j] doesn't have to be used in the best alignment between x[1..i+1] and y[1..j+1]

*

Why 3 Matrices: Example

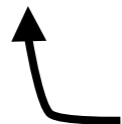match = 5, mismatch = -2, gap = -1, gap_start = -10

x=CARTS, y=CAT

CART
CA-T

OPT(4, 3) = optimal score = 15 - 10 = 5

CARTS
CA-T-

WRONG(5, 3) = 15 - 10 - 10 = -5

CARTS
CAT--

OPT(5, 3) = 10 - 2 - 10 - 1 = -3

this is why we need to keep the X and Y matrices around.
they tell us the score of ending with a gap in one of the sequences.

*

Side Note: Lower Bounds

- Suppose the lengths of *x* and *y* are *n*.

- Clearly, need at least $\Omega(n)$ time to find their global alignment (have to read the strings!)

- The DP algorithms show global alignment can be done in $O(n^2)$ time.

- A trick called the "Four Russians Speedup" can make a similar dynamic programming algorithm run in $O(n^2 / \log n)$ time.

    - We probably won't talk about the Four Russians Speedup.

    - The important thing to remember is that only one of the four authors is Russian…

        (Alrazarov, Dinic, Kronrod, Faradzev, 1970)

- Open questions: Can we do better? Can we prove that we can't do better? No#

#: Backurs, Arturs, and Piotr Indyk. "Edit distance cannot be computed in strongly subquadratic time (unless SETH is false)." *Proceedings of the forty-seventh annual ACM symposium on Theory of computing.* ACM, 2015.

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Recap

- Local alignment: extra "0" case.

- General gap penalties require 3 matrices and $O(n^3)$ time.

- Affine gap penalties require 3 matrices, but only $O(n^2)$ time.

*