

Alignment of short/long-read sequencing data

2019 Dragon Star Bioinformatics Course (Day 2)

Sequence alignment

- Sequence alignment
 - Consider matches, mismatches, insertions and deletions (indels are gaps)
 - Find an optimal alignment between sequences
- Applications
 - Reference-based read mapping
 - Genome assembly
 - Gene finding

Pairwise alignment

- Sequence alignment between two sequences
 - Find an optimal alignment between two sequences
- Dynamic programming
 - Input
 - Two sequences
 - Score matrix

Dynamic programming - score matrix

- Simple scoring matrix
 - Assume: a, b are two bases

- $$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -2 & \text{if } b = \text{gap} \end{cases} \quad \text{or} \quad \phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -2 & \text{for open gaps} \\ -1 & \text{for extended gaps} \end{cases}$$

	A	C	G	T
A	2	-3	-3	-1
C	-2	3	-1	-2
G	-2	-1	4	-3
T	-1	-1	-2	1

Dynamic programming

- Purpose:
 - To find an alignment between two sequences with best matching scores
- Three components
 - Recursive calculation
 - Tabular arrangement
 - Traceback
- Three common types of pairwise alignments
 - Global alignment: Needleman-Wunsch
 - Local alignment: Smith-Waterman
 - Semi-global alignment

Global alignment: Needleman-Wunsch

- Best global alignment
 - Have maximal alignment score with all bases in two sequences
- Assume we have two sequences: P and Q
 - P = TCATGGC
 - Q = TCATC
- Score functions

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

Needleman Wunsch alignment:

- 1. Tabular arrangement
 - $C(0, j) = \sum_{1 \leq k \leq \|P\|} \phi(-, P(k))$
 - $C(i, 0) = \sum_{1 \leq k \leq \|Q\|} \phi(Q(k), -)$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

		<u>Q</u>							
<u>P</u>		-	T	C	A	T	G	G	C
-	-	0	-1	-2	-3	-4	-5	-6	-7
T	T	-1							
C	C	-2							
A	A	-3							
T	T	-4							
C	C	-5							

Global alignment: Needleman-Wunsch

- 2. Recursive calculation $C(i, j) = \max \begin{cases} C(i-1, j-1) + \phi(Q(i), P(j)) \\ C(i-1, j) + \phi(Q(i), -) \\ C(i, j-1) + \phi(-, P(j)) \end{cases}$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

		<u>Q</u>							
<u>P</u>		-	T	C	A	T	G	G	C
	-	0	-1	-2	-3	-4	-5	-6	-7
	T	-1							
	C	-2							
	A	-3							
	T	-4							
	C	-5							

Global alignment: Needleman-Wunsch

- 2. Recursive calculation $C(i, j) = \max \begin{cases} C(i-1, j-1) + \phi(Q(i), P(j)) \\ C(i-1, j) + \phi(Q(i), -) \\ C(i, j-1) + \phi(-, P(j)) \end{cases}$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

		Q							
P		-	T	C	A	T	G	G	C
-	0	-1	-2	-3	-4	-5	-6	-7	
T	-1								
C	-2								
A	-3								
T	-4								
C	-5								

Global alignment: Needleman-Wunsch

- 2. Recursive calculation $C(i, j) = \max \begin{cases} C(i-1, j-1) + \phi(Q(i), P(j)) \\ C(i-1, j) + \phi(Q(i), -) \\ C(i, j-1) + \phi(-, P(j)) \end{cases}$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

		<u>Q</u>							
<u>P</u>		-	T	C	A	T	G	G	C
-	0	-1	-2	-3	-4	-5	-6	-7	
T	-1								
C	-2								
A	-3								
T	-4								
C	-5								

Global alignment: Needleman-Wunsch

- 2. Recursive calculation $C(i, j) = \max \begin{cases} C(i-1, j-1) + \phi(Q(i), P(j)) \\ C(i-1, j) + \phi(Q(i), -) \\ C(i, j-1) + \phi(-, P(j)) \end{cases}$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

<u>Q</u>								
<u>P</u>	-	T	C	A	T	G	G	C
-	0	-1	-2	-3	-4	-5	-6	-7
T	-1	1						
C	-2							
A	-3							
T	-4							
C	-5							

Global alignment: Needleman-Wunsch

- 2. Recursive calculation $C(i, j) = \max \begin{cases} C(i-1, j-1) + \phi(Q(i), P(j)) \\ C(i-1, j) + \phi(Q(i), -) \\ C(i, j-1) + \phi(-, P(j)) \end{cases}$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

Q

<u>P</u>	-	T	C	A	T	G	G	C
-	0	-1	-2	-3	-4	-5	-6	-7
T	-1	1	0	-1	-2	-3	-4	-5
C	-2	0	2	1	0	-1	-2	3
A	-3	-1	1	3	2	1	0	-1
T	-4	-2	-2	0	4	3	2	1
C	-5	-3	1	-1	3	3	2	3

Global alignment: Needleman-Wunsch

- 3. Traceback

- Check which operation obtained the current alignment score.

Q

P	-	T	C	A	T	G	G	C
-	0	-1	-2	-3	-4	-5	-6	-7
T	-1	1	0	-1	-2	-3	-4	-5
C	-2	0	2	1	0	-1	-2	-3
A	-3	-1	1	3	2	1	0	-1
T	-4	-2	-2	0	4	3	2	1
C	-5	-3	-1	-1	3	3	2	3

```

T C A T G G C
| | | |   |
T C A T - - C
  
```

Local alignment: Smith-Waterman

- Best local alignment
 - Have maximal alignment score with a subset of bases in two sequences
- Assume we have two sequences: P and Q
 - P = TCATGGC
 - Q = TCATC
- Score functions

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

Local alignment: Smith-Waterman

- 1. Tabular arrangement

- $C(0, j) = 0$

- $C(i, 0) = 0$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

		<u>Q</u>							
<u>P</u>		-	T	C	A	T	G	G	C
	-	0	0	0	0	0	0	0	0
	T	0							
	C	0							
	A	0							
	T	0							
	C	0							

Local alignment: Smith-Waterman

- 2. Tabular arrangement $(i, j) =$

$$\bullet \max \begin{cases} C(i-1, j-1) + \phi(Q(i), P(j)) \\ C(i-1, j) + \phi(Q(i), -) \\ C(i, j-1) + \phi(-, P(j)) \\ 0 \end{cases}$$

$$\phi(a, b) = \begin{cases} 1 & \text{if } a = b \\ -1 & \text{if } a \neq b \\ -1 & \text{if } a = \text{gap} \\ -1 & \text{if } b = \text{gap} \end{cases}$$

Q
P

	-	T	C	A	T	G	G	C
-	0	0	0	0	0	0	0	0
T	0	1	-1	-1	0	-1	-1	-1
C	0	-1	2	-1	-1	0	-1	1
A	0	-1	-1	3	-1	-1	-1	-1
T	0	-1	-1	-1	4	-1	-1	2
C	0	-1	-1	-1	-1	-1	-1	3

Local alignment: Smith-Waterman

- 3. Traceback

- 1. Find a best score recursively
- 2. Check which operation is used to obtain the current alignment score.
- 3. Stop when an alignment is 0

T C A T
| | | |
T C A T

		<u>Q</u>							
<u>P</u>		-	T	C	A	T	G	G	C
-	-	0	0	0	0	0	0	0	0
T	T	0	1	0	0	1	0	0	0
C	C	0	0	2	1	0	0	0	1
A	A	0	0	1	3	2	1	0	0
T	T	0	0	0	0	4	3	2	1
C	C	0	0	0	0	3	3	2	3

BLAST

- BLAST is used for sequence similarity search, and much faster than Smith-Waterman method
- Compare a query sequence against a database of sequences.
- BLAST is a collection of algorithms
 - BLASTN: nucleotide sequence against nucleotide database
 - BLASTP: protein sequence against protein database
 - BLASTX: translated nucleotide sequence against protein database
 - TBLASTX: six-frame translations of a nucleotide query sequence against the six-frame translations of a nucleotide sequence database
 - TBLASTN: protein sequence against translated nucleotide databases

BLAST

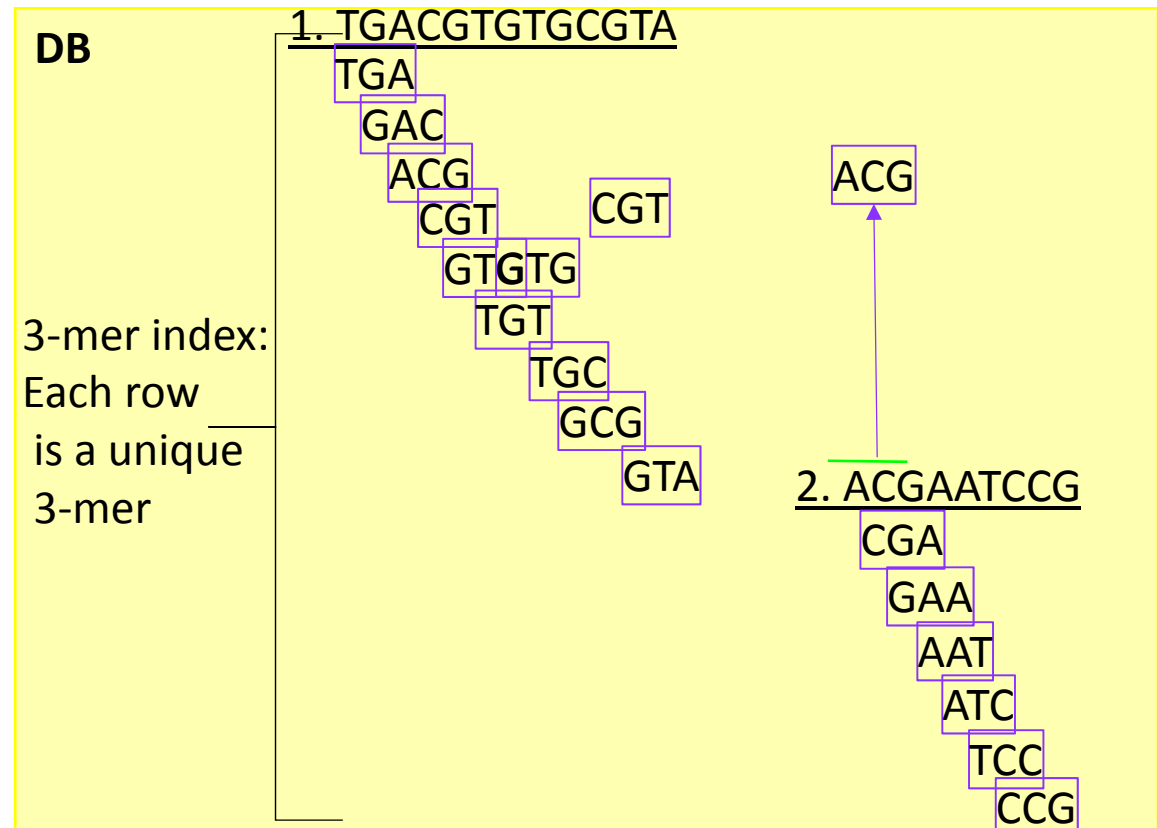
Assume all 3-mer are high-scoring words

- Process: The seed-index-map-extend-merge strategy

- Seed and index:

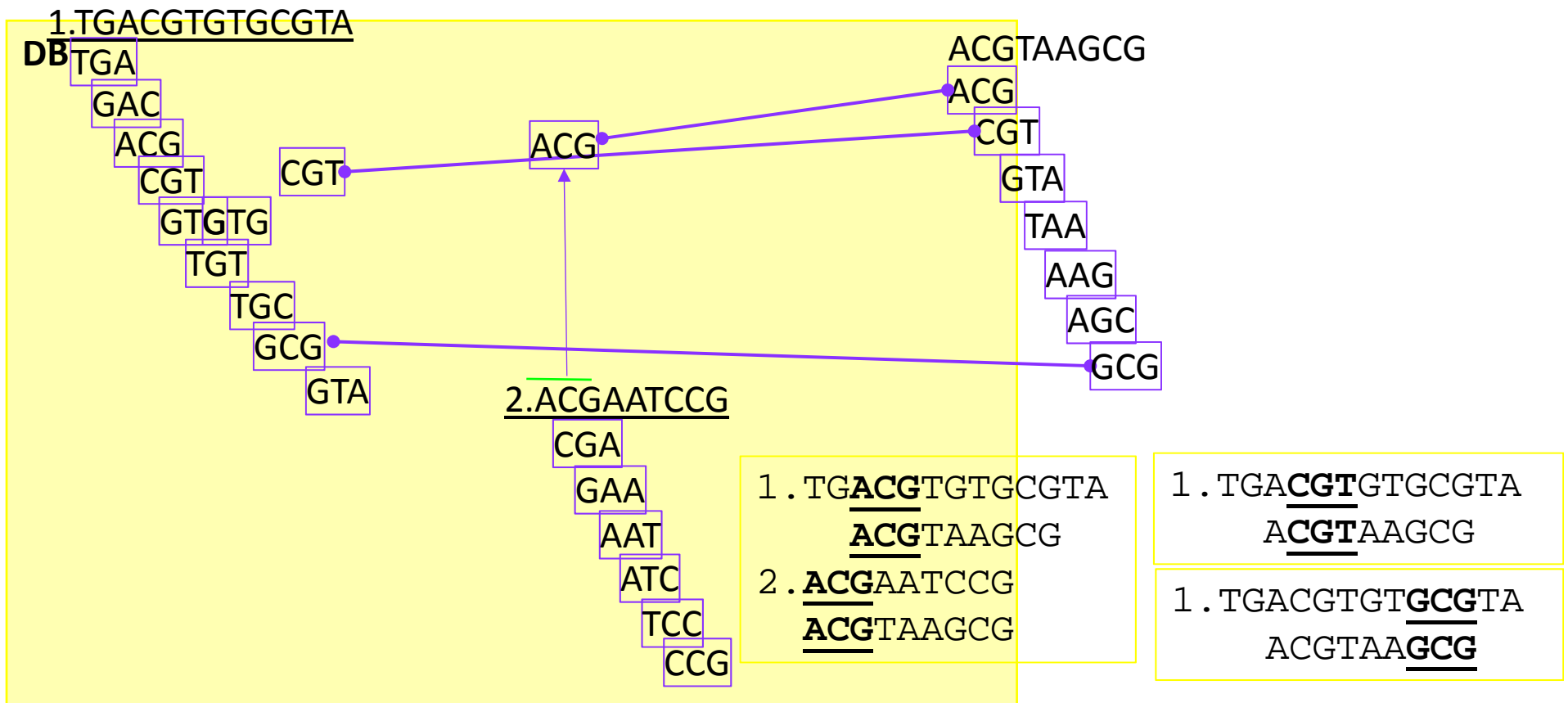
- Construct common words (k-mers) for sequences in a database
- Assume the database has two sequences and k=3
 - 1. TGACGTGTGCGTA
 - 2. ACGAATCCG

- Assume all 3-mers are high-score words
 - BLAST needs a threshold to determine high-score words.



BLAST

- Process: The seed-index-map-extend-merge strategy
 - Seed-map with high-score words
 - Obtain words from a query sequence

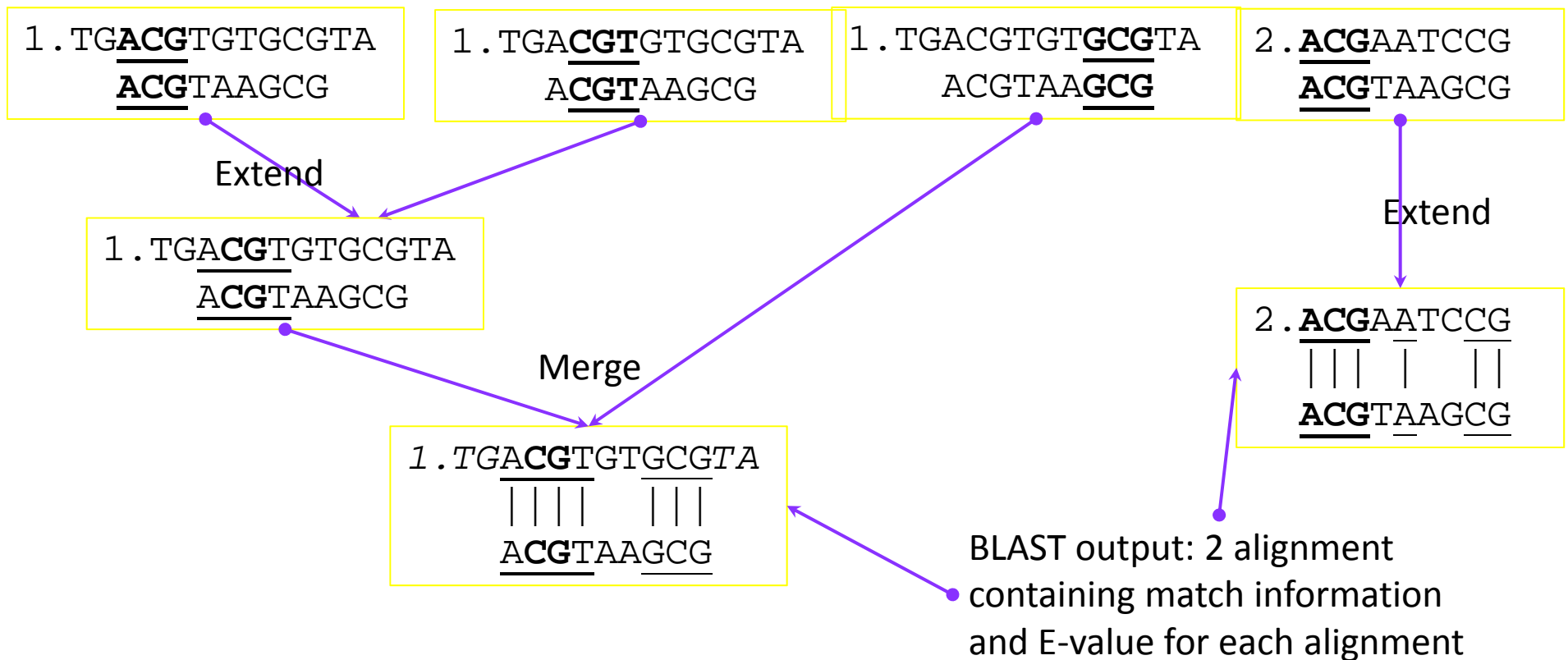


BLAST

- Process: The seed-index-map-extend-merge strategy

- Extend-merge

- Extend: high-scoring segment



Statistical significance of alignment

- How to assess the significance of a high-scoring hit to the database?
 - E is the number of alignments expected by chance during a database search
 - E is a function of the size of the search space (mn), the normalized score (λS), and a constant (K).
- Karlin-Altschul equation: $E = K m n e^{-\lambda S}$
 - Score (S) from scoring matrix
 - K and λ are two constants
 - m : the length of a query sequence
 - n : the sum of bases of all sequences in database
 - The lower E value indicates more significant alignment.
 - $E = -\ln(1 - P)$

Statistical significance of alignment

- How to assess the P-value of finding a high-scoring pair (HSP)?
 - The number of random HSPs with score $\geq S$ is described by a Poisson distribution
 - the probability of finding exactly k HSPs with score $\geq S$ is given by $e^{-E} * E^k/k!$
 - Specifically the chance of finding zero HSPs with score $\geq S$ is e^{-E} , so the probability of finding at least one such HSP is $P=1-e^{-E}$
 - Conversely, $E = -\ln(1 - P)$
 - When E and P are very small, they are almost identical

Bowtie/BWA

- Ultrafast, memory-efficient alignment programs for aligning short DNA sequence reads to large genome
- Burrows–Wheeler transform (BWT)
 - Invented by Burrows and Wheeler, 1994
 - Is a block-sorting compression algorithm for many repeated characters
 - BWA and Bowtie are the most famous implementations of BWT in sequence alignment

Burrows–Wheeler transform (BWT)

- BWT is an invertible transformation of a string S of length n into another string S' of length n
- BWT can organize the genome string into a sequence of suffixes of the original genome string
- Given a string S , BWT
 - Add $\$$ and assume $\$ <$ any alphabet
 - Obtains a suffix array of all cyclic rotations
 - Transform into a new string S'

BWT procedures

- Given a string $S = \text{"TCATC"}$, BWT
 - Adds $\$$ and assume $\$ < \text{any alphabet}$
 - Obtains a suffix array of all cyclic rotations

Transformation				
1. Input	2. Cyclic rotations	3. Sorting	4. Last column	5. BWT output
TCATC\$				

BWT procedures

- Cyclic rotation:
 - Obtains a suffix array of all cyclic rotations
 - Keeps an index of the rotated strings in the array
 - Creates Circular Permutation Table (CPT)

Transformation				
1. Input	2. Cyclic rotations	3. Sorting	4. Last column	5. BWT output
TCATC\$	0 TCATC\$ 1 CATC\$T 2 ATC\$TC 3 TC\$TCA 4 C\$TCAT 5 \$TCATC			

BWT procedures

- Sorting:
 - Sorts the Circular Permutation Table (CPT) alphabetically
 - Keep the index with the strings

Transformation				
1. Input	2. Cyclic rotations	3. Sorting	4. Last column	5. BWT output
TCATC\$	0 TCATC\$ 1 CATC\$T 2 ATC\$TC 3 TC\$TCA 4 C\$TCAT 5 \$TCATC	5 \$TCATC 2 ATC\$TC 4 C\$TCAT 1 CATC\$T 3 TC\$TCA 0 TCATC\$		

BWT procedures

- Taking the last column from the sorted array

Transformation				
1. Input	2. Cyclic rotations	3. Sorting	4. Last column	5. BWT output
TCATC\$	0 TCATC\$ 1 CATC\$T 2 ATC\$TC 3 TC\$TCA 4 C\$TCAT 5 \$TCATC	5 \$TCATC 2 ATC\$TC 4 C\$TCAT 1 CATC\$T 3 TC\$TCA 0 TCATC\$	5 \$TCAT C 2 ATC\$T C 4 C\$TCA T 1 CATC\$ T 3 TC\$TCA A 0 TCATC \$	

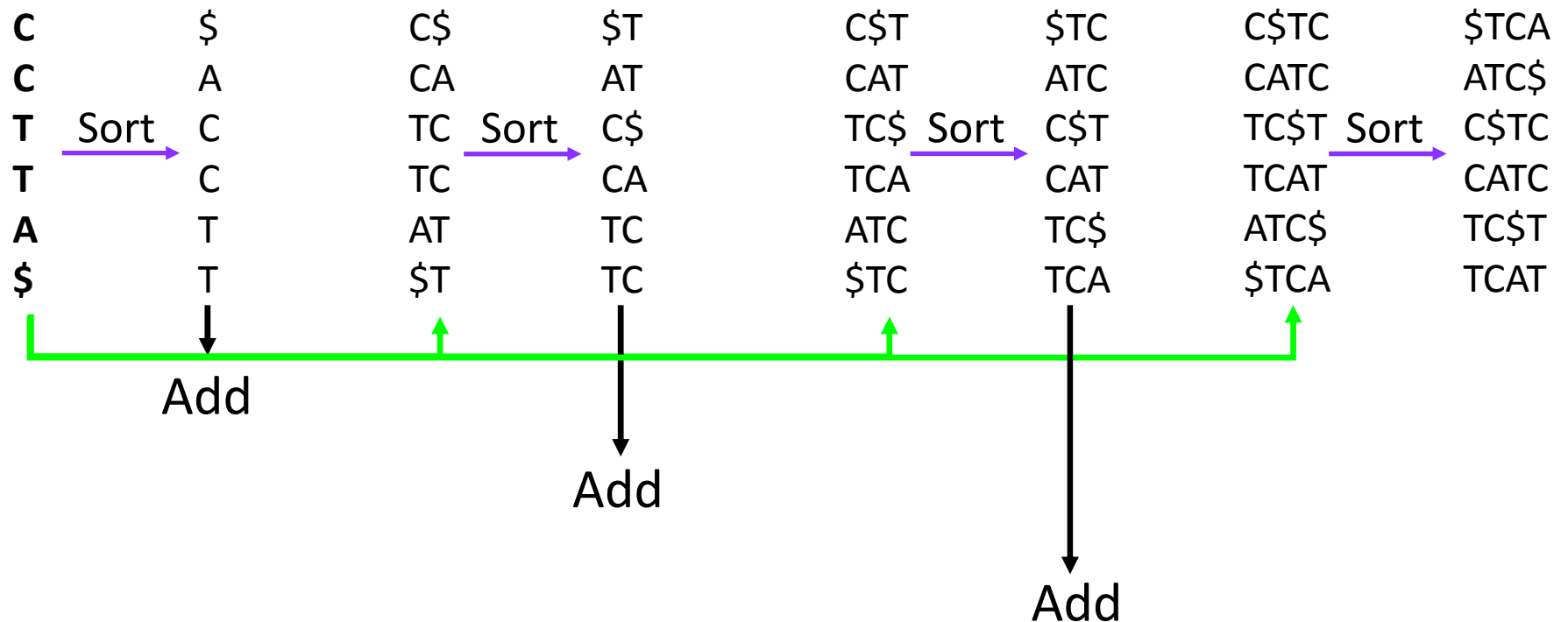
BWT procedures

- BWT output:
 - The last column represents the transformed string

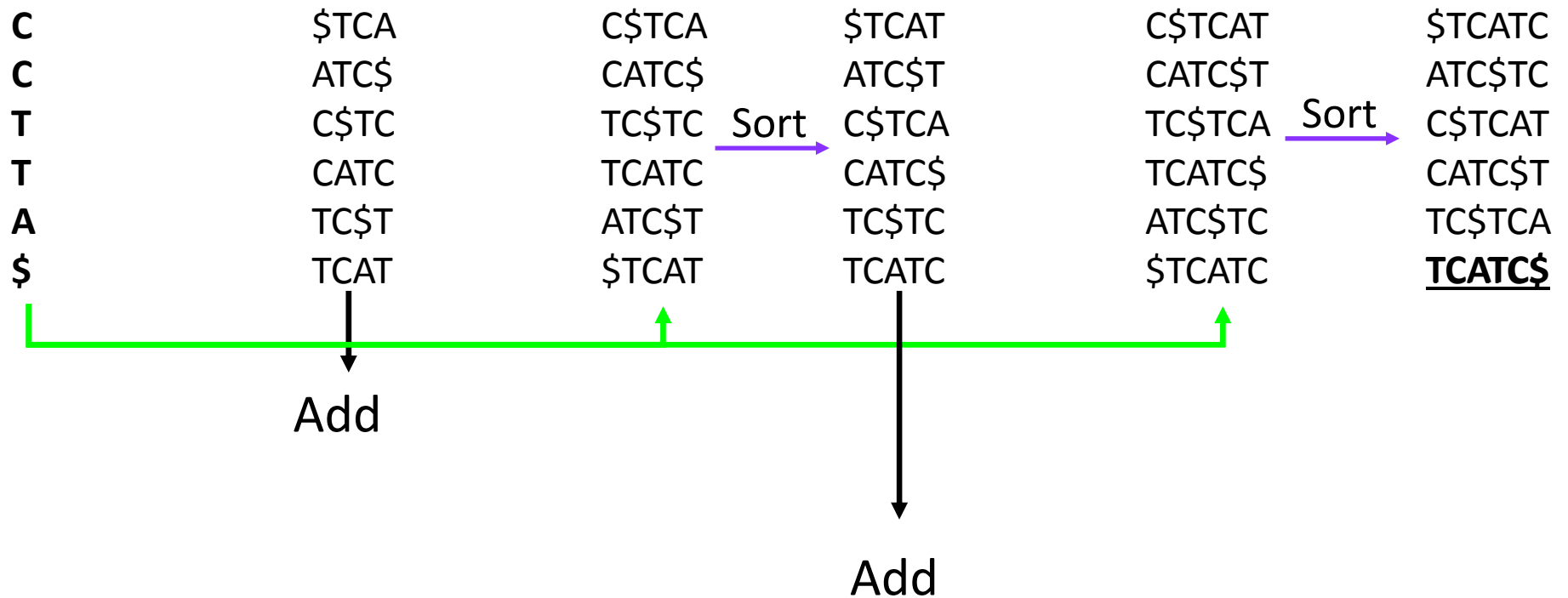
Transformation				
1. Input	2. Cyclic rotations	3. Sorting	4. Last column	5. BWT output
TCATC\$	0 TCATC\$ 1 CATC\$T 2 ATC\$TC 3 TC\$TCA 4 C\$TCAT 5 \$TCATC	5 \$TCATC 2 ATC\$TC 4 C\$TCAT 1 CATC\$T 3 TC\$TCA 0 TCATC\$	5 \$TCAT C 2 ATC\$T C 4 C\$TCA T 1 CATC\$ T 3 TC\$TCA A 0 TCATC \$	CCTTA\$

BWT reverse transformation

- Transform back to the original string
 - With a suffix array, it is easy to recover the original string S
 - Input: CCTTA\$ for TCATC\$

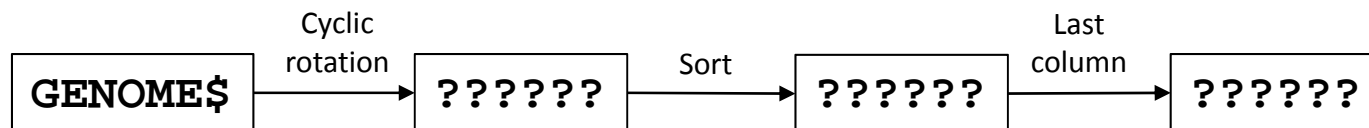


BWT reverse transformation



What's the BWT for GENOME\$

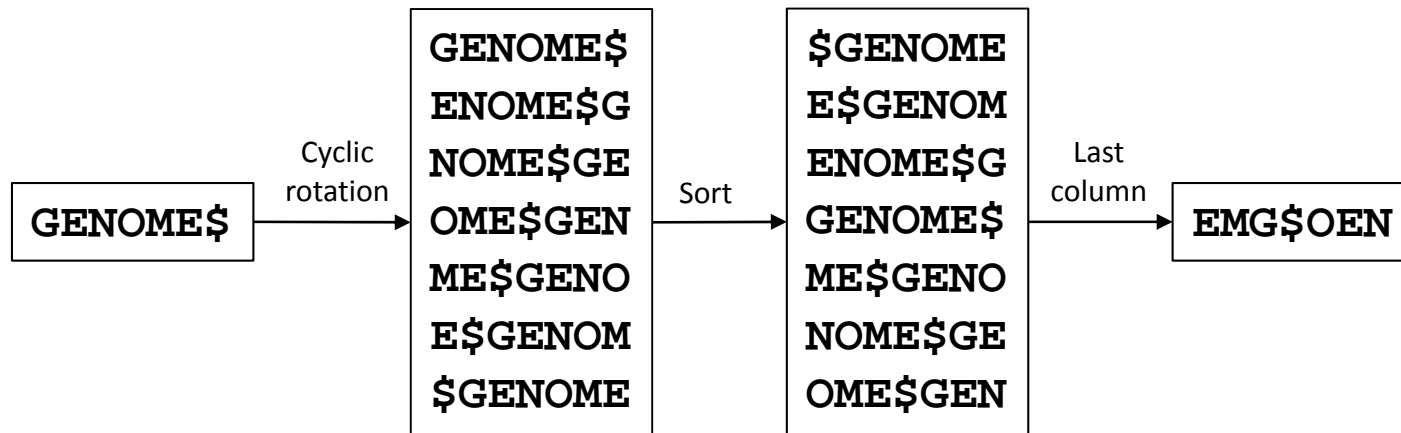
BWT (GENOME\$) = ?



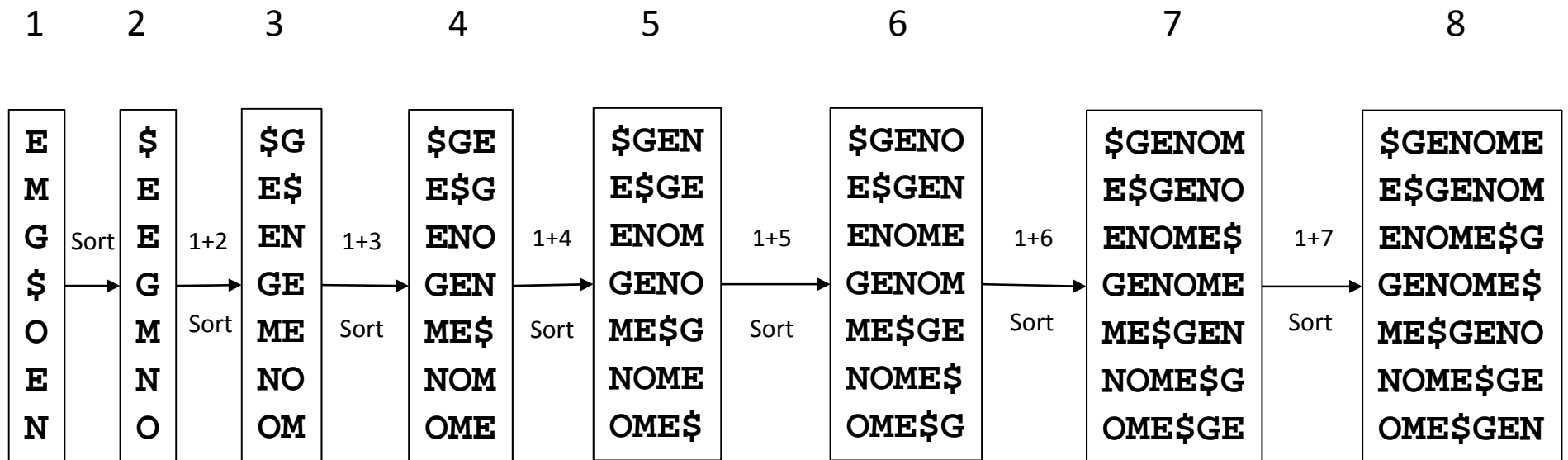
After class exercise: reproduce the BWT transformation above

What's the BWT for GENOME?

BWT (GENOME\$) = EMG\$OEN



BWT reverse transformation



After class exercise: reproduce the reverse transformation above

BWT search

- Match process:
 - Given a query string, we want to find where is the matched sub-string in original string: TCATC\$
 - We do this search through the use of \mathcal{LF} functions on the BWT of the original string
 - \mathcal{LF} (last to first) functions:
 - Purpose: find the same base in the first column corresponding to a specific base in the last column of BWT matrix

LF property

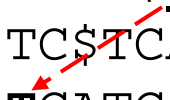
- \mathcal{LF} is a function mapping from the last column of the BWT matrix to the first column
- The i th occurrence of the character X in the last column of BWT matrix corresponds to the i th occurrence of X in the first column.
- Examples:
 - The 1st occurrence of C in the last column is the same C as the 1st occurrence of C in the first column
 - The 2nd occurrence of T in the last column is the same T as the 2nd occurrence of T in the first column

```
$TCATC
ATC$TC
C$TCAT
CATC$T
TC$TCA
TCATC$
```

LF calculation

- $LF(pos, na) = C(na) + pr_{na}(pos, na)$
 - pos : position in the output CCTTA\$
 - na : a base
 - $C(na)$: the number of bases smaller than na in BWT first column
 - $pr_{na}(pos, na)$: the number of base na before the position of pos (not included).
- For TCATC (see right side), the $C()$ for \$, A, C, T are 0, 1, 2, 4, respectively. This is pre-computed.
- For example, $LF(3, 'T')$
 - $C('T') = 4$, $pr_{na}(3, 'T') = 1$
 - Then: $LF(3, 'T') = 4+1 = 5$
 - The two underlined T in bold (one in the last column and one in the first column) as the same T in the original string.

0	\$TCATC
1	ATC\$TC
2	C\$TCAT
3	CATC\$ <u>T</u>
4	TC\$TCA
5	<u>T</u> CATC\$



T in the last column $pos = 3$ is the same T in first column $pos = 5$

Now, you can try do the same exercise on the C in the last column $pos = 1$

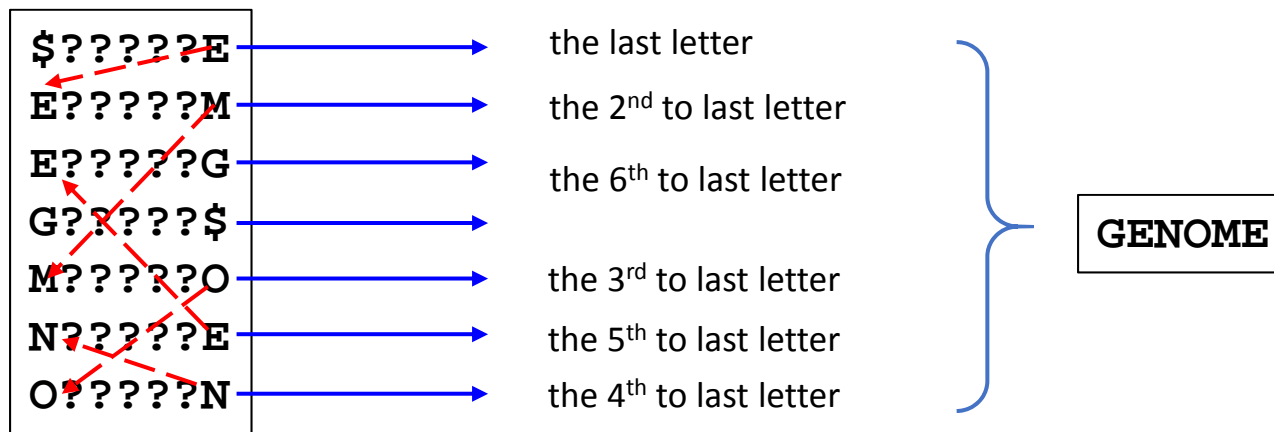
How LF recovers original string

- Suppose we only know first/last column of BWT, what is the original string?

\$?	?	?	?	?	E
E	?	?	?	?	?	M
E	?	?	?	?	?	G
G	?	?	?	?	?	\$
M	?	?	?	?	?	O
N	?	?	?	?	?	E
O	?	?	?	?	?	N

How LF recovers original string

- Suppose we only know first/last column of BWT, what is the original string?



BWT search

- Purpose: search a specific pattern (query) in the BWT matrix, and recover the index in original database sequence

- Example: search “AT” within the database sequence “TCATC” (the index of “AT” is 2)

0	1	2	3	4	5
T	C	A	T	C	\$

- Next we define the lower and upper bound of search recursively (from last to first base of query)
 - $L(W)$: lowest index in BWT matrix where W is prefix
 - $U(W)$: highest index in BWT matrix where W is prefix
 - The $L()$ and $U()$ for A, C, G, T can be pre-calculated from the BWT matrix
 - For a new prefix p in front of W , we will have:
 - $L(pW) = \text{LF}(L(W), p)$
 - $U(pW) = \text{LF}(U(W)+1, p)-1$

BWT search

- Match process:

- Input BWT: CCTTA\$
- Query: AT

C	C	T	T	A	\$
0	1	2	3	4	5

- The match process is:

- 1. for 'T' in "AT"

- $L(T) = 4$
- $U(T) = 5$

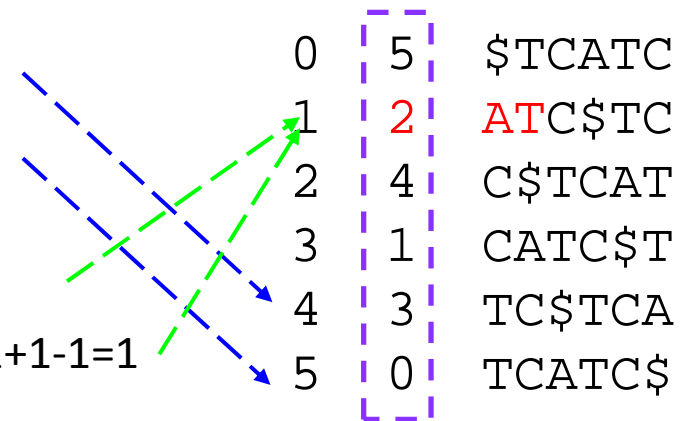
- 2. for 'A' in "AT"

- $L(AT) = \text{LF}(L(T), 'A') = \text{LF}(4, 'A') = 1 + 0 = 1$
- $U(AT) = \text{LF}(U(T) + 1, 'A') - 1 = \text{LF}(6, 'A') - 1 = 1 + 1 - 1 = 1$

- The matched row is: 1

- The index in original string is 2

0	1	2	3	4	5
T	C	A	T	C	\$



0	5	\$TCATC
1	2	ATC\$TC
2	4	C\$TCAT
3	1	CATC\$T
4	3	TC\$TCA
5	0	TCATC\$

BWT search

- Match process:

- Input BWT: CCTTA\$
- Query: TC

C	C	T	T	A	\$
0	1	2	3	4	5

- The match process is:

- 1. for 'C' in "TC"

- $L(C) = 2$
- $U(C) = 3$

- 2. for 'T' in "TC"

- $L(TC) = \text{LF}(L(C), 'T') = \text{LF}(2, 'T') = 4 + 0 = 4$
- $U(TC) = \text{LF}(U(C) + 1, 'T') - 1 = \text{LF}(4, 'T') - 1 = 4 + 2 - 1 = 5$

- The matched row is: 4 and 5.

- The index in original string is 0 and 3.

0	5	\$TCATC
1	2	ATC\$TC
2	4	C\$TCAT
3	1	CATC\$T
4	3	TC\$TCA
5	0	TCATC\$

0 1 2 3 4 5
T C A T C \$

BWT search

- Match process:

- Input BWT: CCTTA\$
- Query: AC

C	C	T	T	A	\$
0	1	2	3	4	5

- The match process is:

- 1. for 'C' in "AC"

- $L(C) = 2$
- $U(C) = 3$

- 2. for 'A' in "AC"

- $L(AC) = \text{LF}(L(C), 'A') = \text{LF}(2, 'A') = 1 + 0 = 1$
- $U(AC) = \text{LF}(U(C) + 1, 'A') - 1 = \text{LF}(4, 'A') - 1 = 1 + 0 - 1 = 0$

- The matched row is None.

- "AC" is not in "TCATC"

0	5	\$TCATC
1	2	ATC\$TC
2	4	C\$TCAT
3	1	CATC\$T
4	3	TC\$TCA
5	0	TCATC\$

BWT search

- Match process:

- Input reference: CCTTA\$
- Query: ATC

C	C	T	T	A	\$
0	1	2	3	4	5

- The match process is:

- 1. for 'C' in "ATC"
 - $L(C) = 2$
 - $U(C) = 3$
- 2. for 'T' in "ATC"
 - $L(TC) = \text{LF}(L(C), 'T') = \text{LF}(2, 'T') = 4 + 0 = 4$
 - $U(TC) = \text{LF}(U(C) + 1, 'T') - 1 = \text{LF}(4, 'T') - 1 = 4 + 2 - 1 = 5$
- 3. for "A" in "ATC"
 - $L(ATC) = \text{LF}(4, 'A') = 1 + 0 = 1$
 - $U(ATC) = \text{LF}(6, 'A') - 1 = 1 + 1 - 1 = 1$

0	5	\$TCATC
1	2	ATC\$TC
2	4	C\$TCAT
3	1	CATC\$T
4	3	TC\$TCA
5	0	TCATC\$

- The matched row is 1

- The original index is 2

0	1	2	3	4	5
T	C	A	T	C	\$

Minimap2

- Fast and accurate aligner for whole genome sequencing data against a reference genome
 - Can work with both short reads and noisy long reads
 - A typical seed-chain-alignment strategy
 - 1. Collect minimizers of reference sequence
 - 2. Index in a hash table
 - 3. Get minimizers from query sequences
 - 4. Find exact match as anchors
 - 5. Find collinear anchors
 - 6. Extend or close gaps by dynamic programming

Traditional k-mers methods vs minimap2

- Traditional k-mer based sequence similarity
 - BLAST:
 - Get k-mer, generate hash value and store in a hash table
 - Find k-mer match between reference sequence and query sequence
 - DALIGNER:
 - Generate k-mer for each of two sets of reads
 - Sort k-mers and merge them for potential match
 - MHAP:
 - m k-mer hash functions
 - For each hash functions, find minimum hash value for all k-mers in a sequence
 - Two sequences are similar to each other if they have many overlaps of minimum hash value
- Minimap uses hash functions, but only focuses on “minimizers”, to reduce storage and improve speed substantially

The concept of “minimizer”

- Chooses a representative k-mer from a group of adjacent k-mers, so different strings T_i and T_j choose the same representative if they share a long enough subsequence.
- Only a small fraction of k-mers, called ‘minimizers’, needs to be stored.

Position	1	2	3	4	5	6	7	1	2	3	4	5	6	7	8	9	10	11	12
Sequence	2	3	1	0	3	4	3	4	2	6	4	7	2	8	1	4	7	5	1
<i>k</i> -mers	2	3	1					4	2	6	4	7	2	8					
with		3	1	0					2	6	4	7	2	8	1				
minimizer			1	0	3					6	4	7	2	8	1	4			
in				0	3	4					4	7	2	8	1	4	7		
bold					3	4	3					7	2	8	1	4	7	5	
	(a)							(b)					2	8	1	4	7	5	1

Minimap2

- 1. Collect minimizers of reference sequence

- Minimizer:

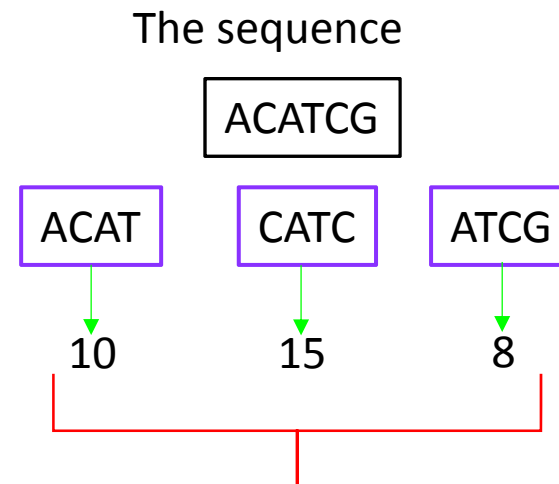
- A smallest k -mer of w consecutive k -mers in a sequence with $w+k$ bases
 - Minimap2: $k=15$ and $w=5$

- Example:

- Assume $w=3$ and $k=4$

three k -mer (4-mer)

hash value



The minimizer of ACATCG is 8

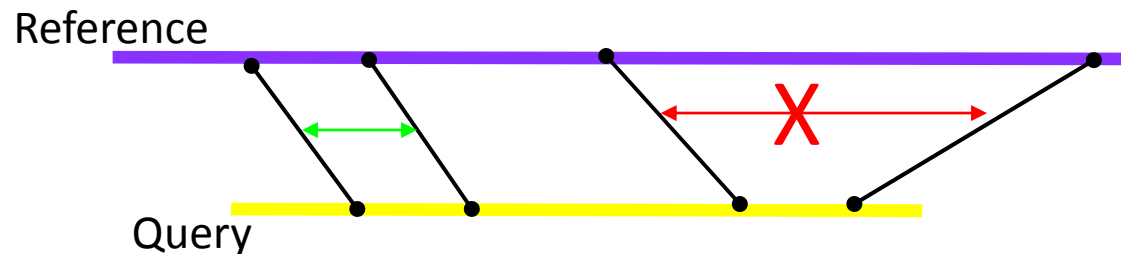
- Given a sequence with $>w+k$ bases
 - A set of minimizers were generated

Minimap2

- 2. Index in a hash table
 - Key: a hash value of a minimizer
 - Values: list of locations of minimizer in the reference genome
- 3. Get minimizers from query sequences
 - Used the same hash function to obtain a set of minimizers for query sequence
- 4. Find exact match as anchors
 - Match minimizers between reference and query sequence
 - The matched sub-sequence of reference and sub-sequence of query: anchors

Minimap2

- 5. Find collinear anchors
 - Two anchors are collinear
 - Their distance in a sequence is less than a threshold.
 - Forward and reverse strands are considered individually
 - Cluster close anchors.



- 6. Extend or close gaps between anchors by dynamic programming