



Advanced Bioinformatics

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Day 2: Genome Replication (Part 2) 26th Feb. 2024





Outlines

- Introduction to Bio Data analysis (Python)
- Computational Application in Genome Replication
- Genome Replication Problem (Part 2)
 - Asymmetry of Replication, Skew Diagrams, Finding Frequent Words with Mismatches
- Bioinformatics Challenges with using python







Outlines

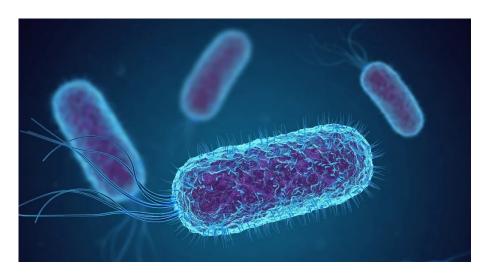
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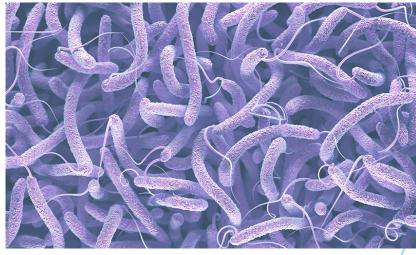




Metagenomics Example





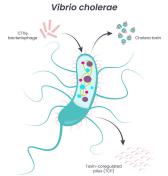




For example : Vibrio cholerae

OriC of VC







atcaatgatcaacgtaagcttctaagcATGATCAAGgtgctcacacagtttatccacaac cggaaagATGATCAAGagaggatgatttcttggccatatcgcaatgaatacttgtgactt gtgcttccaattgacatcttcagcgccatattgcgctggccaaggtgacggagcgggatt acgaaagcatgatcatggctgttgttctgtttatcttgttttgactgagacttgttagga tagacggtttttcatcactgactagccaaagccttactctgcctgacatcgaccgtaaat tgataatgaatttacatgcttccgcgacgatttacctcttgatcatcgatccgattgaag atcttcaattgttaattctcttgcctcgactcatagccatgatgagctcttgatcatgtt tccttaaccctctattttttacggaagaATGATCAAGctgctgctcttgatcatcgtttc



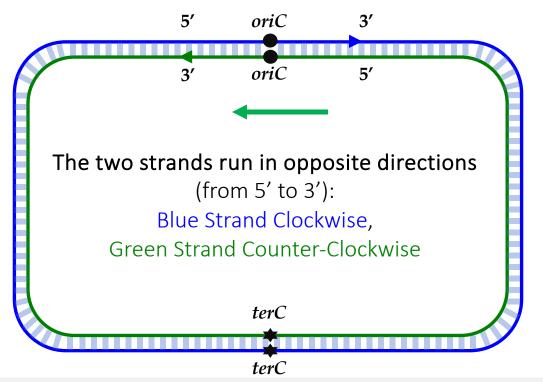
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DNA Strands Have Directions!





As the strands unwind, they create two **replication forks**, which expand in both directions around the chromosome until the strands completely separate at the **replication terminus** (denoted *ter*).









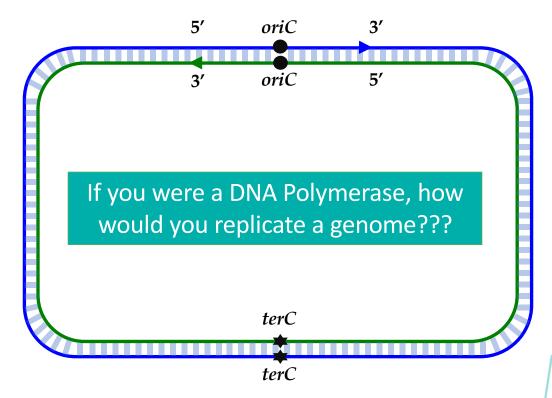
Think and share your thoughts:)





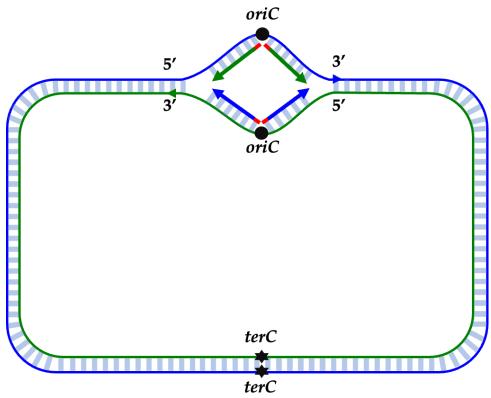
DNA Strands Have Directions







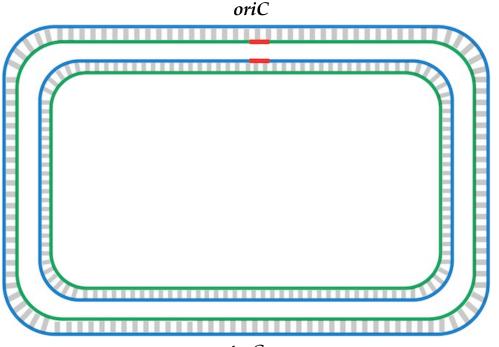
Four DNA Polymerases Do the Job





Four DNA Polymerases Do the Job



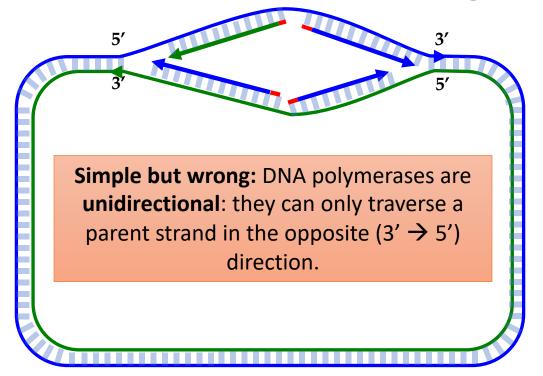


terC

When all four DNA polymerases have reached *ter*, the chromosome's DNA will have been completely replicated, resulting in two pairs of complementary strands shown in the lower figure, and the cell is ready to divide.

Continue as Replication Fork Enlarges





DNA polymerases are **unidirectional**, meaning that they can only traverse a template strand of DNA in the $3' \rightarrow 5'$ direction, which is opposite from the $5' \rightarrow 3'$ direction of DNA.



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Forward and Backward Strands

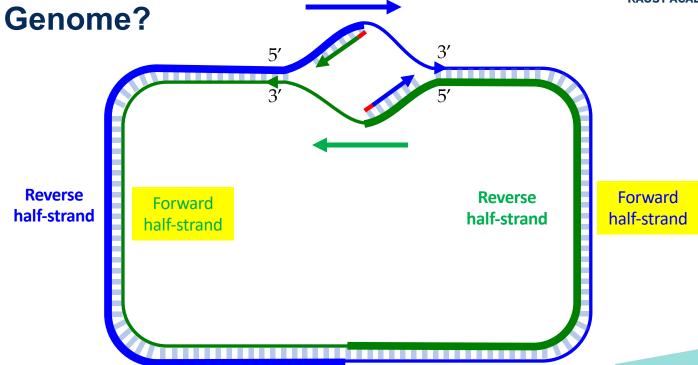


- The unidirectionality of DNA polymerase requires a major revision to our naive model of replication.
- There are four different half-strands of parent DNA connecting oriC to terC, as highlighted in the following Figure.
- Two of these half-strands are traversed from oriC to terC in the 5' -> 3'
 direction and are thus called forward half-strands.



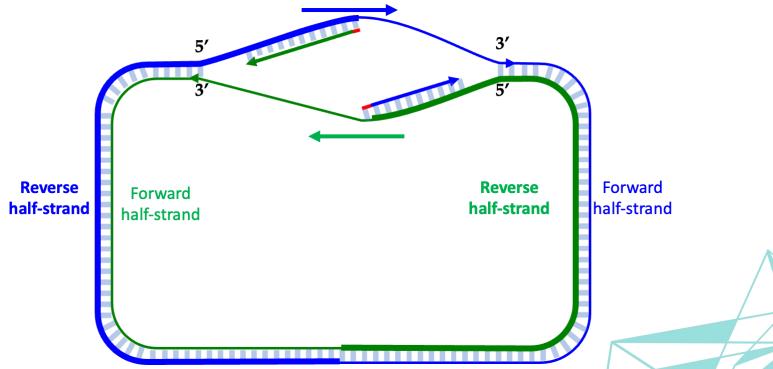
If you Were a UNIDIRECTIONAL DNA Polymerase, how Would you Replicate a







If you Were a UNIDIRECTIONAL DNA Polymerase, How Would you Replicate a Genome???





Asymmetry of Replication

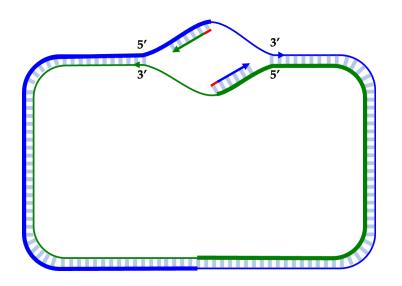


- Since a DNA polymerase can only move in the reverse (3' → 5') direction, it can copy nucleotides non-stop from oriC to terC along reverse half-strands.
- However, replication on forward half-strands is very different because a DNA polymerase cannot move in the forward (5' → 3') direction; on these half-strands, a DNA polymerase must replicate backwards toward oriC.
- DNA polymerase must wait for the replication fork to open a little (approximately 2,000 nucleotides) until a new primer is formed at the end of the replication fork; afterwards, the DNA polymerase starts replicating a small chunk of DNA starting from this primer and moving backward in the direction of oriC.



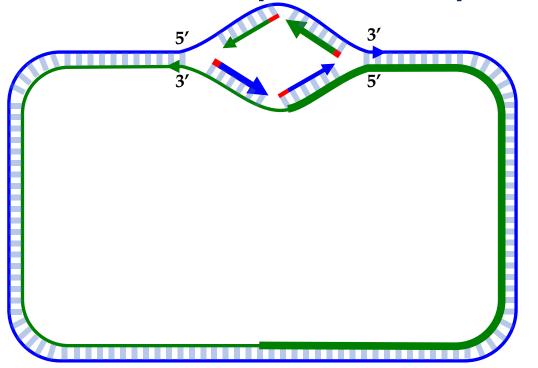
Wait until the Fork Opens and...





On a forward half-strand, in order to replicate DNA, a DNA polymerase must wait for the replication fork to open a little (approximately 2,000 nucleotides) until a new primer is formed at the *end* of the replication fork; afterwards, the DNA polymerase starts replicating a small chunk of DNA starting from this primer and moving *backward* in the direction of *ori*. When the two DNA polymerases on forward half-strands reach *ori*, we have the situation shown below.

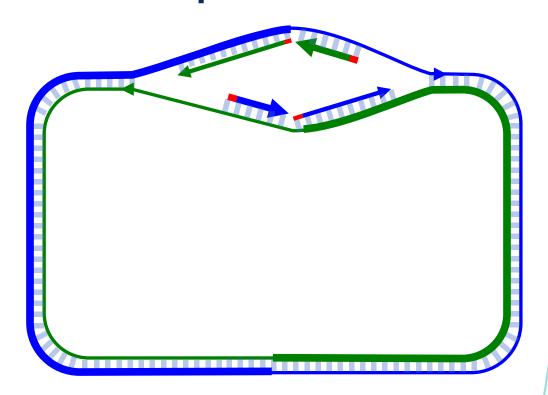
Wait until the Fork Opens and Replicate



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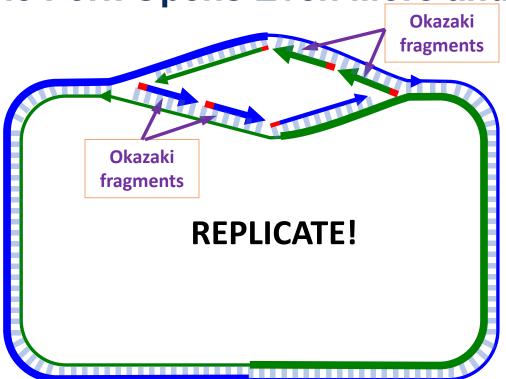
On the whole, replication on a forward half-strand requires occasional stopping and restarting, which results in the synthesis of short **Okazaki fragments** from multiple primers that are complementary to intervals on the forward half-strands

Wait until the Fork Opens and Replicate Wait until the Fork Opens Even More and KAUST ACADEMY





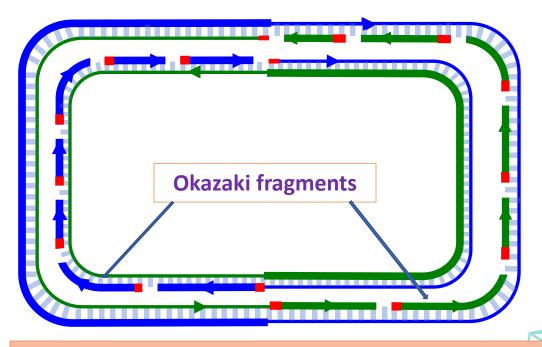
Wait until the Fork Opens and Replicate Wait until the Fork Opens Even More and..





Okazaki Fragments Need to be Ligated to Fill in the Gaps

Biologists call a reverse half-strand (thick lines) a leading half-strand since a single DNA polymerase traverses this half-strand nonstop, and they call a forward half-strand (thin lines) a lagging half-strand since it is used as a template by many DNA polymerases stopping and starting replication.



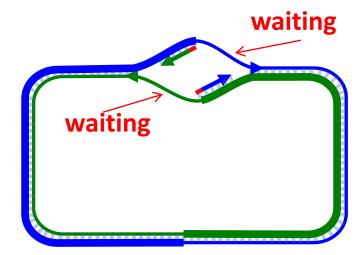


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Different Lifestyles of Reverse and Forward Half-Strands



- The reverse half-strand lives a doublestranded life most of the time.
- The forward half-strand spends a large portion of its life single-stranded, waiting to be replicated.



But why would a computer scientist care?





Asymmetry of Replication Affects Nucleotide Frequencies

Single-stranded DNA has a much higher mutation rate than double-stranded DNA.

Thus, if one nucleotide has a greater mutation rate, then we should observe its **shortage** on the forward half-strand that lives single-stranded life!

Which nucleotide (A/C/G/T) has the highest mutation rate? Why?





The Peculiar Statistics of #G - #C

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Cytosine (C) rapidly mutates into thymine (T) through **deamination**; deamination rates rise 100-fold when DNA is single stranded!

Forward half-strand (single-stranded life): **shortage of C, normal G**Reverse half-strand (double-stranded life): **shortage of G, normal C**

Difference

#C #G #G - #C
Reverse half-strand 219518 201634 -17884
Forward half-strand 207901 211607 +3706

+11617

-9973

deamination:

Spontaneous deamination con verts cytosine to uracil



Deamination

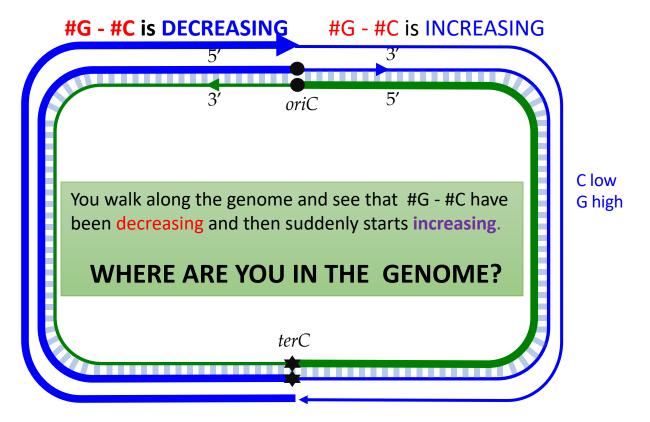


| | #C | #G | # A | $\#\mathbf{T}$ |
|----------------------|--------|--------|------------|----------------|
| Entire strand | 427419 | 413241 | 491488 | 491363 |
| Reverse half-strand | 219518 | 201634 | 243963 | 246641 |
| Forward half-strand | 207901 | 211607 | 247525 | 244722 |
| Difference | +11617 | -9973 | -3562 | +1919 |

STOP and Think: Do you notice anything about the nucleotide counts in this table?



Take a Walk Along the Genome



C high/G low → #G - #C is DECREASING as we walk along the REVERSE half-strand

C high

G low

Let's see if we

advantage of these peculiar

can take

statistics

caused by

locate ori.

deamination to

C low/G high \rightarrow #G - #C is INCREASING as we walk along the FORWARD half-strand



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Bioinformatics Challenges



Finding Frequent Words with Mismatches



- 1F: Minimize Skew
- 1G: Hamming Distance Between Two Strings
- 1H: Approximate Occurrences of a Pattern
- 1J: The Most Frequent Mismatches



Bioinformatics Challenges



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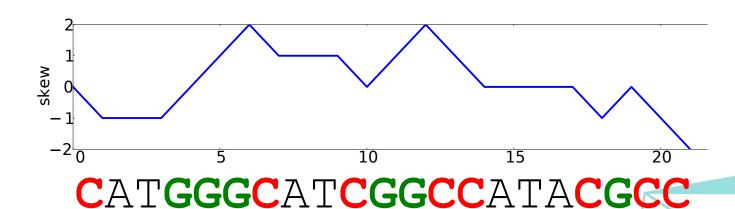


Skew Diagram



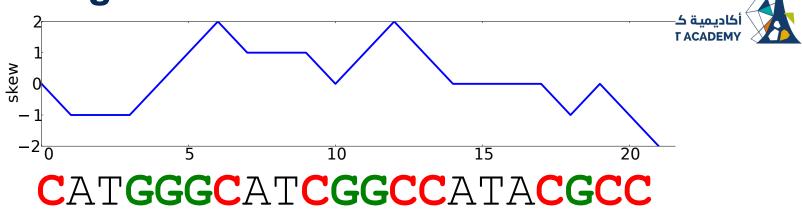
Skew(k): #G - #C for the first k nucleotides of Genome.

Skew diagram: Plot *Skew(k)* against *k*





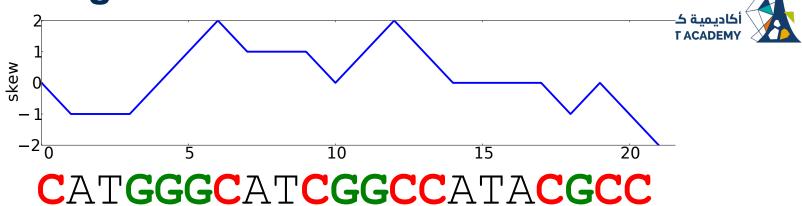
Skew Diagram



Since we don't know the location of *ori* in a circular genome, let's linearize it (i.e., select an arbitrary position and pretend that the genome begins here), resulting in a linear string *Genome*. We define *Skew_i(Genome)* as the difference between the total number of occurrences of G and the total number of occurrences of C in the first *i* nucleotides of *Genome*.



Skew Diagram



The **skew diagram** is defined by plotting $Skew_i(Genome)$ (as i ranges from 0 to |Genome|), where $Skew_0(Genome)$ is set equal to zero. The figure below shows a skew diagram for the DNA string CATGGGCATCGGCCATACGCC. Note that we can compute $Skew_{i+1}(Genome)$ from $Skew_i(Genome)$ according to the nucleotide in position i of Genome. If this nucleotide is G, then $Skew_{i+1}(Genome) = Skew_i(Genome) + 1$; if this nucleotide is G, then $Skew_{i+1}(Genome) = Skew_i(Genome) = Skew_i(Genome) = Skew_i(Genome)$.



Minimum Skew Problem



Minimum Skew Problem:

Find a position in a genome where the skew diagram attains a minimum.

Input: A DNA string *Genome*.

Output: All integer(s) i minimizing $SKEW_i(Genome)$ among all values of i

(from 0 to |Genome|).





Bioinformatics Challenges



Finding Frequent Words with Mismatches



- 1F: Minimize Skew
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Hamming Distance



For X & Y where |X| = |Y|, hamming distance = minimum # substitutions needed to turn one into the other

X: GAGGTAGCGGCGTTTAAC

Y: GTGGTAACGGGGTTTAAC





Hamming Distance



For X & Y where |X| = |Y|, hamming distance = minimum # substitutions needed to turn one into the other





2. Hamming Distance



Hamming Distance Problem:

Compute the Hamming distance between two strings.

Input: Two strings of equal length.

Output: The Hamming distance between these strings.





Bioinformatics Challenges





- 1F: Minimize Skew
- 1G: Hamming Distance Between Two Strings
- 1H: Approximate Occurrences of a Pattern
- 1J: The Most Frequent Mismatches



1. The Approximate Pattern Matching Problem.



```
ApproximatePatternCount(Text, Pattern, d)
    count ← 0
    for i \leftarrow 0 to |Text| - |Pattern|
        Pattern' ← Text(i , |Pattern|)
        if HammingDistance(Pattern, Pattern') ≤ d
             count ← count + 1
    return count
```

Code Challenge: Implement ApproximatePatternCount()



Bioinformatics Challenges





1F: Minimize Skew

1G: Hamming Distance Between Two Strings

1H: Approximate Occurrences of a Pattern

• 1J: The Most Frequent Mismatches



Frequent Words with Mismatches Problem

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Frequent Words with Mismatches Problem: *Find the most frequent k-mers with mismatches in a string.*

- •Input: A string *Text* as well as integers *k* and *d*.
- •Output: All most frequent *k*-mers with up to *d* mismatches in *Text*.

For example, to generate Neighbors(CAA,1), first form Neighbors(AA,1) = {AA, CA, GA, TA, AC, AG, AT}. The Hamming distance between AA and each of six of these neighbors is 1. Firstly, concatenating C with each of these patterns results in seven patterns (CAA, CCA, CGA, CTA, CAC, CAG, and CAT) that belong to Neighbors(CAA, 1). Secondly, concatenating any nucleotide with AA results in four patterns (AAA, CAA, GAA, and TAA) that belong to Neighbors(CAA, 1). Thus, Neighbors(CAA, 1) comprises eleven patterns.



Frequent Words with Mismatches Problem

```
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```

```
FrequentWordsWithMismatches(Text, k, d)
    Patterns ← an array of strings of length 0
   freqMap ← empty map
    n ← |Text|
    for i \leftarrow 0 to n - k
        Pattern \leftarrow Text(i, k)
        neighborhood ← Neighbors(Pattern, d)
        for j \leftarrow 0 to |neighborhood| - 1
            neighbor ← neighborhood[j]
            if freqMap[neighbor] doesn't exist
                freaMap[neighbor] ← 1
            else
                freqMap[neighbor] + freqMap[neighbor] + 1
    m ← MaxMap(freqMap)
    for every key Pattern in fregMap
        if freqMap[Pattern] = m
            append Pattern to Patterns
    return Patterns
```





Frequent Words with Mismatches and Reverse Complements Problem.



Frequent Words with Mismatches and Reverse Complements Problem: Find the most frequent k-mers (with mismatches and reverse complements) in a string.

- •Input: A DNA string *Text* as well as integers *k* and *d*.
- •Output: All k-mers Pattern maximizing the sum $Count_d(Text, Pattern) + Count_d(Text, Pattern_{rc})$ over all possible k-mers.







Done with Day 2, Heyyyyy!

Thank You!



