
CSE 527 Computational Biology Autumn 2006

Lectures 2-3
Sequence Alignment;
DNA Replication

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This week

- Sequence alignment
- More sequence alignment
- Weekly “bio” interlude - DNA replication

Sequence Alignment

Part I
Motivation, dynamic programming,
global alignment

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Sequence Alignment

- What
- Why
- A Simple Algorithm
- Complexity Analysis
- A better Algorithm:
“Dynamic Programming”

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Sequence Similarity: What

```

GGACCA
TACTAAG
TCCAAT
    
```

Sequence Similarity: What

```

GGACCA
TACTAAG
| : | : | :
TCC-AAT
    
```

Sequence Similarity: Why

- Most widely used comp. tools in biology
- New sequence always compared to sequence data bases

Similar sequences often have similar origin or function

- Selection operates on system level, but mutation occurs at the sequence level
- Recognizable similarity after $10^8 - 10^9$ yr

BLAST Demo

<http://www.ncbi.nlm.nih.gov/blast/>

Taxonomy Report

```

root ..... 64 hits 16 orgs
. Eukaryota ..... 62 hits 14 orgs [cellular organisms]
. . Fungi/Metazoa group ..... 57 hits 11 orgs
. . . Bilateria ..... 38 hits 7 orgs [Metazoa; Eumetazoa]
. . . . Coelomata ..... 36 hits 6 orgs
. . . . . Tetrapoda ..... 26 hits 5 orgs [;;; Vertebrata;;; Sarcopterygii]
. . . . . Eutheria ..... 24 hits 4 orgs [Amniota; Mammalia; Theria]
. . . . . Homo sapiens ..... 20 hits 1 orgs [Primates;; Hominidae; Homo]
. . . . . Murinae ..... 3 hits 2 orgs [Rodentia; Sciurognathi; Muridae]
. . . . . Rattus norvegicus ..... 2 hits 1 orgs [Rattus]
. . . . . Mus musculus ..... 1 hits 1 orgs [Mus]
. . . . . Sus scrofa ..... 1 hits 1 orgs [Cetartiodactyla; Suina; Suidae; Sus]
. . . . . Xenopus laevis ..... 2 hits 1 orgs [Amphibia;;;; Xenopodinae; Xenopus]
. . . . . Drosophila melanogaster ..... 10 hits 1 orgs [Protostomia;;; Drosophila;;]
. . . . . Caenorhabditis elegans ..... 2 hits 1 orgs [; Nematoda;;;;; Caenorhabditis]
. . . . . Ascomycota ..... 19 hits 4 orgs [Fungi]
. . . . . Schizosaccharomyces pombe ..... 10 hits 1 orgs [;;; Schizosaccharomycetes]
. . . . . Saccharomycetales ..... 9 hits 3 orgs [Saccharomycotina; Saccharomycetes]
. . . . . Saccharomyces ..... 8 hits 2 orgs [Saccharomycetaceae]
. . . . . Saccharomyces cerevisiae ..... 7 hits 1 orgs
. . . . . Saccharomyces kluyveri ..... 1 hits 1 orgs
. . . . . Candida albicans ..... 1 hits 1 orgs [mitosporic Saccharomycetales;]
. . . . . Arabidopsis thaliana ..... 2 hits 1 orgs [Viridiplantae; Brassicaceae;]
. . . . . Apicomplexa ..... 3 hits 2 orgs [Alveolata]
. . . . . Plasmodium falciparum ..... 2 hits 1 orgs [Haemosporida; Plasmodium]
. . . . . Toxoplasma gondii ..... 1 hits 1 orgs [Coccidia; Eimeriida; Sarcocystidae;]
. . . . . synthetic construct ..... 1 hits 1 orgs [other; artificial sequence]
. . . . . 1mhbhovev1+ disease virus ..... 1 hits 1 orgs [Viruses; dsRNA viruses, no RNA ...]
    
```

Terminology (CS, not necessarily Bio)

- **String**: ordered list of letters TATAAG
- **Prefix**: consecutive letters from front
empty, T, TA, TAT, ...
- **Suffix**: ... from end
empty, G, AG, AAG, ...
- **Substring**: ... from ends or middle
empty, TAT, AA, ...
- **Subsequence**: ordered, nonconsecutive
TT, AAA, TAG, ...

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Sequence Alignment

```

a c b c d b      a c - - b c d b
 c' a d b d      - c a d b - d -
  
```

- Defn:** An **alignment** of strings S, T is a pair of strings S', T' (with spaces) s.t.
- (1) $|S'| = |T'|$, and $(|S| = \text{"length of S"})$
 - (2) removing all spaces leaves S, T

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Alignment Scoring

Mismatch = -1
Match = 2

```

a c b c d b      a c - - b c d b
c a d b d        - c a d b - d -
                -1 2  -1 -1 2  -1 2  -1
Value = 3*2 + 5*(-1) = +1
  
```

- The **score** of aligning (characters or spaces) x & y is $\sigma(x,y)$.
- **Value** of an alignment $\sum_{i=1}^{|S'|} \sigma(S'[i], T'[i])$
- An **optimal alignment**: one of max value

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Optimal Alignment: A Simple Algorithm

- for all** subseqs A of S, B of T s.t. $|A| = |B|$ **do**
 align A[i] with B[i], $1 \leq i \leq |A|$
 align all other chars to spaces
 compute its value
 retain the max
end
 output the retained alignment

```

S = abcd  A = cd
T = wxyz  B = xz
-abc-d   a-bc-d
w--xyz   -w-xyz
  
```

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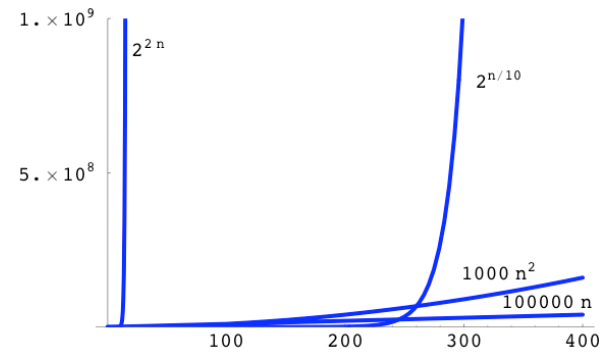
Analysis

- Assume $|S| = |T| = n$
- Cost of evaluating one alignment: $\geq n$
- How many alignments are there: $\geq \binom{2n}{n}$
 pick n chars of S, T together
 say k of them are in S
 match these k to the k unpicked chars of T
- Total time: $\geq n \binom{2n}{n} > 2^{2n}$, for $n > 3$
- E.g., for $n = 20$, time is $> 2^{40}$ operations

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Polynomial vs Exponential Growth



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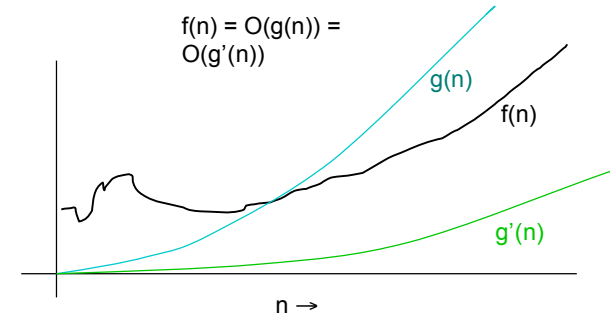
Asymptotic Analysis

- How does run time grow as a function of problem size?
 n^2 or $100n^2 + 100n + 100$ vs 2^{2n}
- **Defn:** $f(n) = O(g(n))$ iff there is a constant c s.t. $|f(n)| \leq cg(n)$ for all sufficiently large n .
 $100n^2 + 100n + 100 = O(n^2)$ [e.g. $c = 101$]
 $n^2 = O(2^{2n})$
 2^{2n} is *not* $O(n^2)$

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Big-O Example



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Utility of Asymptotics

- “All things being equal,” smaller asymptotic growth rate is better
- All things are never equal
- Even so, big-O bounds often let you quickly pick most promising candidates among competing algorithms
- Poly time algs often practical; non-poly algs seldom are.

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Fibonacci Numbers

```
fib(n) {  
  if (n <= 1) {  
    return 1;  
  } else {  
    return fib(n-1) + fib(n-2);  
  }  
}
```

Simple recursion,
but many
repeated
subproblems!!
=>
Time = $\Omega(1.61^n)$

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Fibonacci, II

```
int fib[n];  
fib[0] = 1;  
fib[1] = 1;  
for(i=2; i<=n; i++) {  
  fib[i] = fib[i-1] + fib[i-2];  
}  
return fib[n];
```

Avoid repeated
subproblems by
tabulating them
=>
Time = $O(n)$

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Candidate for Dynamic Programming?

- Common Subproblems?
 - Plausible: probably re-considering alignments of various small substrings unless we're careful.
- Optimal Substructure?
 - Plausible: left and right "halves" of an optimal alignment probably should be optimally aligned (though they obviously interact a bit at the interface).
- (Both made rigorous below.)

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Optimal Substructure (In More Detail)

- Optimal alignment ends in 1 of 3 ways:
 - last chars of S & T aligned with each other
 - last char of S aligned with space in T
 - last char of T aligned with space in S
 - (never align space with space; $\sigma(-, -) < 0$)
- In each case, the **rest** of S & T should be optimally aligned to each other

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Optimal Alignment in $O(n^2)$ via “Dynamic Programming”

- Input: S, T, $|S| = n$, $|T| = m$
- Output: **value** of optimal alignment

Easier to solve a “harder” problem:

$V(i,j)$ = value of optimal alignment of
 $S[1], \dots, S[i]$ with $T[1], \dots, T[j]$
 for **all** $0 \leq i \leq n$, $0 \leq j \leq m$.

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Base Cases

- $V(i,0)$: first i chars of S all match spaces

$$V(i,0) = \sum_{k=1}^i \sigma(S[k], -)$$

- $V(0,j)$: first j chars of T all match spaces

$$V(0,j) = \sum_{k=1}^j \sigma(-, T[k])$$

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General Case

Opt align of $S[1], \dots, S[i]$ vs $T[1], \dots, T[j]$:

$$\begin{matrix} \text{Opt align of} \\ S_1 \dots S_{i-1} \text{ \& } \\ T_1 \dots T_{j-1} \end{matrix} \left[\begin{array}{c} \text{~~~~} S[i] \\ \text{~~~~} T[j] \end{array} \right], \left[\begin{array}{c} \text{~~~~} S[i] \\ \text{~~~~} - \end{array} \right], \text{ or } \left[\begin{array}{c} \text{~~~~} - \\ \text{~~~~} T[j] \end{array} \right]$$

$$V(i,j) = \max \left\{ \begin{array}{l} V(i-1,j-1) + \sigma(S[i], T[j]) \\ V(i-1,j) + \sigma(S[i], -) \\ V(i,j-1) + \sigma(-, T[j]) \end{array} \right\},$$

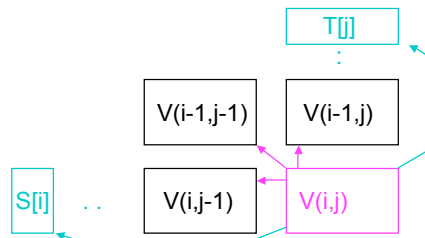
for all $1 \leq i \leq n$, $1 \leq j \leq m$.

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Calculating One Entry

$$V(i,j) = \max \begin{cases} V(i-1,j-1) + \sigma(S[i], T[j]) \\ V(i-1,j) + \sigma(S[i], -) \\ V(i,j-1) + \sigma(-, T[j]) \end{cases}$$



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Example

Mismatch = -1
Match = 2

j	0	1	2	3	4	5	
i			c	a	d	b	d
0	0	-1	-2	-3	-4	-5	
1	a	-1	-1	1			
2	c	-2	1				
3	b	-3					
4	c	-4					
5	d	-5					
6	b	-6					

← T

Time = O(mn)

↑ S

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Example

Mismatch = -1
Match = 2

j	0	1	2	3	4	5	
i			c	a	d	b	d
0	0	-1	-2	-3	-4	-5	
1	a	-1	-1	1	0	-1	-2
2	c	-2	1	0	0	-1	-2
3	b	-3	0	0	-1	2	1
4	c	-4	-1	-1	-1	1	1
5	d	-5	-2	-2	1	0	3
6	b	-6	-3	-3	0	3	2

← T

↑ S

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Finding Alignments: Trace Back

j	0	1	2	3	4	5	
i			c	a	d	b	d
0	0	-1	-2	-3	-4	-5	
1	a	-1	-1	1	0	-1	-2
2	c	-2	1	0	0	-1	-2
3	b	-3	0	0	-1	2	1
4	c	-4	-1	-1	-1	1	1
5	d	-5	-2	-2	1	0	3
6	b	-6	-3	-3	0	3	2

← T

↑ S

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Complexity Notes

- Time = $O(mn)$, (value and alignment)
- Space = $O(mn)$
- Easy to get **value** in Time = $O(mn)$ and Space = $O(\min(m,n))$
- Possible to get value **and alignment** in Time = $O(mn)$ and Space = $O(\min(m,n))$ but tricky.

Sequence Alignment

Part II Local alignments & gaps

Variations

- Local Alignment
 - Preceding gives *global* alignment, i.e. full length of both strings;
 - Might well miss strong similarity of part of strings amidst dissimilar flanks
- Gap Penalties
 - 10 adjacent spaces cost 10 x one space?
- Many others

Local Alignment: Motivations

- “Interesting” (evolutionarily conserved, functionally related) segments may be a small part of the whole
 - “Active site” of a protein
 - Scattered genes or exons amidst “junk”, e.g. retroviral insertions, large deletions
 - Don’t have whole sequence
- Global alignment might miss them if flanking junk outweighs similar regions

Local Alignment

Optimal *local alignment* of strings S & T:
Find substrings A of S and B of T
having max value global alignment

S = abcxdex A = c x d e
T = xxxcde B = c - d e value = 5

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The “Obvious” Local Alignment Algorithm

for all substrings A of S and B of T
 Align A & B via dynamic programming
 Retain pair with max value
end ;
Output the retained pair

Time: $O(n^2)$ choices for A, $O(m^2)$ for B,
 $O(nm)$ for DP, so $O(n^3m^3)$ total.

[Best possible? Lots of redundant work...]

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Local Alignment in $O(nm)$ via Dynamic Programming

- Input: S, T, $|S| = n$, $|T| = m$
- Output: value of optimal *local alignment*

Better to solve a “harder” problem
for all $0 \leq i \leq n$, $0 \leq j \leq m$:

$V(i,j)$ = **max** value of opt (global)
alignment of a **suffix** of $S[1], \dots, S[i]$
with a **suffix** of $T[1], \dots, T[j]$

Report best i,j

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Base Cases

- Assume $\sigma(x,-) \leq 0$, $\sigma(-,x) \leq 0$
- $V(i,0)$: some suffix of first i chars of S; all
match spaces in T; best suffix is empty
 $V(i,0) = 0$
- $V(0,j)$: similar
 $V(0,j) = 0$

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General Case Recurrences

Opt suffix align $S[1], \dots, S[i]$ vs $T[1], \dots, T[j]$:

$$\begin{bmatrix} \text{~~~~~} S[i] \\ \text{~~~~~} T[j] \end{bmatrix}, \begin{bmatrix} \text{~~~~~} S[i] \\ \text{~~~~~} - \end{bmatrix}, \begin{bmatrix} \text{~~~~~} - \\ \text{~~~~~} T[j] \end{bmatrix}, \text{ or } \begin{bmatrix} \text{~~~~~} \\ \text{~~~~~} \end{bmatrix}$$

Opt align of
suffix of
 $S_1 \dots S_{i-1}$ &
 $T_1 \dots T_{j-1}$

$$V(i,j) = \max \begin{cases} V(i-1,j-1) + \sigma(S[i], T[j]) \\ V(i-1,j) + \sigma(S[i], -) \\ V(i,j-1) + \sigma(-, T[j]) \\ 0 \end{cases}$$

opt suffix alignment has:
2, 1, 1, 0
chars of S/T

for all $1 \leq i \leq n, 1 \leq j \leq m$.

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Scoring Local Alignments

	j	0	1	2	3	4	5	6	
i			x	x	x	c	d	e	←T
0		0	0	0	0	0	0	0	
1	a	0							
2	b	0							
3	c	0							
4	x	0							
5	d	0							
6	e	0							
7	x	0							↑S

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Finding Local Alignments

	j	0	1	2	3	4	5	6	
i			x	x	x	c	d	e	←T
0		0	0	0	0	0	0	0	
1	a	0	0	0	0	0	0	0	
2	b	0	0	0	0	0	0	0	
3	c	0	0	0	0	2	1	0	
4	x	0	2	2	2	1	1	0	
5	d	0	1	1	1	1	3	2	
6	e	0	0	0	0	0	2	5	
7	x	0	2	2	2	1	1	4	↑S

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Notes

- Time and Space = $O(mn)$
- Space $O(\min(m,n))$ possible with time $O(mn)$, but finding alignment is trickier
- Local alignment: "Smith-Waterman"
- Global alignment: "Needleman-Wunsch"

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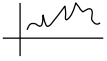

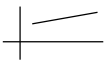
Alignment With Gap Penalties

- **Gap:** maximal run of spaces in S' or T'
 ab----c-d
 a-ddddcbd 2 gaps in S', 1 in T'
- Motivations, e.g.:
 - mutation might insert/delete several or even many residues at once
 - matching cDNA (no introns) to genomic DNA (exons and introns)

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Gap Penalties

- Score = $f(\text{gap length})$
- Kinds, & best known alignment time
 - general  $O(n^3)$
 - convex  $O(n^2 \log n)$
 - affine  $O(mn)$

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Global Alignment with Affine Gap Penalties

$V(i,j)$ = value of opt alignment of
 $S[1], \dots, S[i]$ with $T[1], \dots, T[j]$
 $G(i,j)$ = ..., s.t. last pair matches $S[i]$ & $T[j]$
 $F(i,j)$ = ..., s.t. last pair matches $S[i]$ & -
 $E(i,j)$ = ..., s.t. last pair matches - & $T[j]$

Time: $O(mn)$ [calculate all, $O(1)$ each]

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Affine Gap Algorithm

Gap penalty = $g + s * (\text{gap length})$, $g, s \geq 0$
 $V(i,0) = E(i,0) = V(0,i) = F(0,i) = -g - i * s$
 $V(i,j) = \max(G(i,j), F(i,j), E(i,j))$
 $G(i,j) = V(i-1, j-1) + \sigma(S[i], T[j])$
 $F(i,j) = \max(\underbrace{F(i-1, j) - s}_{\text{old gap}}, \underbrace{V(i-1, j) - g - s}_{\text{new gap}})$
 $E(i,j) = \max(\underbrace{E(i, j-1) - s}_{\text{old gap}}, \underbrace{V(i, j-1) - g - s}_{\text{new gap}})$

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Summary

- Functionally similar proteins/DNA often have recognizably similar sequences even after eons of divergent evolution
- Ability to find/compare/experiment with “same” sequence in other organisms is a huge win
- Surprisingly simple scoring model works well in practice: score each position separately & add, possibly w/ fancier gap model like affine
- Simple “dynamic programming” algorithms can find *optimal* alignments under these assumptions in poly time (product of sequence lengths)
- This, and heuristic approximations to it like BLAST, are workhorse tools in molecular biology

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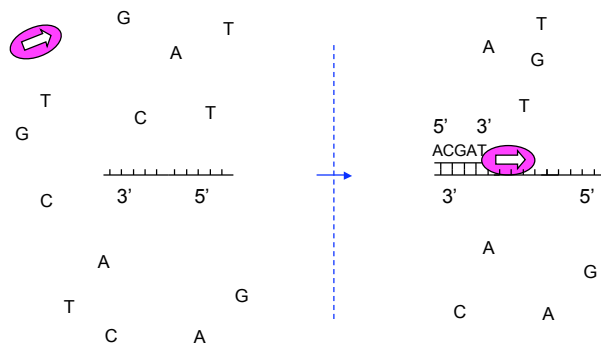
Weekly Bio Interlude

DNA Replication

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DNA Replication: Basics

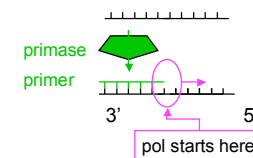


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Issues & Complications, I

- 1st ~10 nt's added are called the *primer*
- In simple model, DNA pol has 2 jobs: prime & extend
- Priming is error-prone
- So, specialized *primase* does the priming; pol specialized for fast, accurate extension
- Still doesn't solve the accuracy problem (hint: primase makes an *RNA* primer)

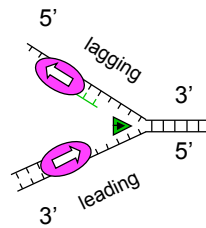


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Issue 2: Rep Forks & Helices

- “Replication Fork”: DNA double helix is progressively unwound by a DNA **helicase**, and both resulting single strands are duplicated
- DNA **polymerase** synthesizes new strand 5' → 3' (reading its template strand 3' → 5')
- That means on one (the “leading”) strand, DNA pol is chasing/pushing the replication fork
- But on the other “lagging” strand, DNA pol is running away from it.

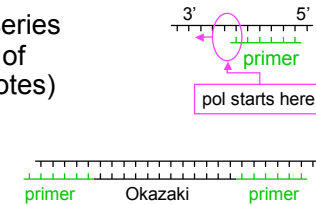


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Issue 3: Fragments

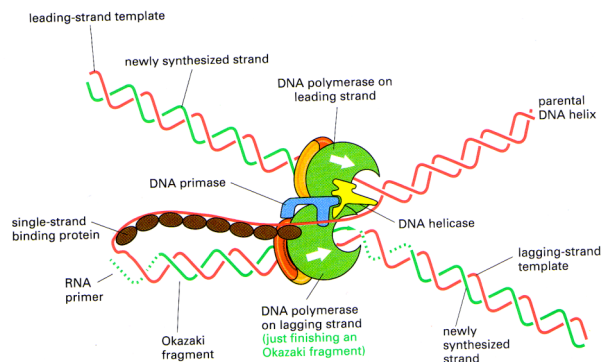
- Lagging strand gets a series of “Okazaki fragments” of DNA (~200nt in eukaryotes) following each primer
- The RNA primers are later removed by a **nuclease** and DNA pol fills gaps (more accurate than primase)
- Fragments joined by **ligase**



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Issue 4: Coord Lead/Lag



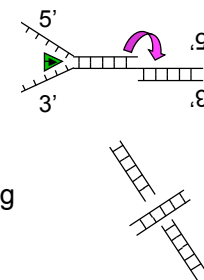
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Alberts et al., Mol. Biol. of the Cell, 3rd ed, p258

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Issue 5: Twirls & Tangles

- Unwinding helix (~10 nucleotides per turn) would cause stress. **Topoisomerase I** cuts DNA backbone on *one* strand, allowing it to spin about the remaining bond, relieving stress
- **Topoisomerase II** can cut & rejoin *both* strands, after allowing another double strand to pass through the gap, de-tangling it.



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Issue 6: Proofreading

- Error rate of pol itself is $\sim 10^{-4}$, but overall rate is 10^{-9} , due to proofreading & repair, e.g.
 - pol itself can back up & cut off a mismatched base if one happens to be inserted
 - priming the new strand is hard to do accurately, hence RNA primers, later removed & replaced
 - other enzymes scan helix for “bulges” caused by base mismatch, figure out which strand is original, cut away new (faulty) copy; DNA pol fills gap
 - which strand is original? In bacteria, some A's are “methylated”, but not immediately after replication

Replication Summary

- Speed: 50 (eukaryotes) - 500 (prokaryotes) bp/sec
- Accuracy: 1 error per 10^9 bp
- Complex & highly optimized
- Highly similar across all living cells

- More info:
Alberts et al., Mol. Biol. of the Cell