BMEG3105 Fall 2023

Data Analytics for Personalized Genomics and Precision Medicine

Lecture 4: Dynamic Programming

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Lecture outline:

- Recap from last lecture
- Dynamic Programming

Part 1. Recap from last lecture

Sequence Data:

- DNA sequence
  - Composed of A, T, C, G bases
  - Consists of complementary double strands
- RNA sequence:
  - Composed of A, U, C, G bases
- Protein sequence:
  - Composed of 20 amino acids
  - Multiple sequence alignment
- To find the best pairwise alignment
  - With two sequences and scoring matrix known
  - Enumeration as a solution
    - Enumerate all the possible alignments between two sequences
    - Calculate scores for all alignments
    - Choose the alignment with the highest score
    - Problem with this solution is that there are too many possible alignments

Part 2. Dynamic Programming (DP):

- Break down a problem into smaller sub-problems
- Solve these sub-problems optimally and recursively
- Use these optimal solutions to construct the optimal solution for the original problem

Flight problem example:

- Direct flight not always cheap
- Finite pre-destination
- Break the flight into several flight trips
- Compare the sum of flight trips for each option and choose the cheapest option

Cheapest (KAUST, CUHK) = Cheapest [Cheapest (KAUST, Qatar) + Cheapest (Qatar, CUHK) ; Cheapest (KAUST, Dubai) + Cheapest (Dubai, CUHK) ; Cheapest (KAUST, GZ) + Cheapest (GZ, CUHK)]

Similarly with Optimal alignment score:

- Each base either aligns to a gap or another base
- Alignment score equals to the sum of score for each alignment pair

Sequence alignment with DP:

- Consider the possibilities of the last pair of the alignment
- Calculate alignment score and choose the best option
- Break down the best option and repeat the last step

ACCG and ACG alignment example:

1. Consider the last base pair. (_ represent blank)

F(ACCG, ACG) = Best [F(ACC, ACG) + F(G, _) ; F(ACCG, AC) + F(_, G) ; F(ACC, AC) + S(G, G)]

F(ACCG, ACG) = Best [F(ACC, ACG) – 10 ; F(ACCG, AC) – 10 ; F(ACC, AC) + 2]

Scoring matrix:

	A	C	G	T
A	2	-7	-5	-7
C	-7	2	-7	-5
G	-5	-7	2	-7
T	-7	-5	-7	2

Gap penalty = -10

➔ Choose the option with highest last base pair score, i.e. F(ACC, AC) + 2

2. Break down unknown part the option chosen

F(ACC, AC) = Best [F(AC, AC)+ F(C, _) ; F(ACC, A) + F(_, C) = Best [F(AC, AC) – 10 ; F(ACC, A) – 10 ; F(AC, A) +2] ; F(AC, A) + S(C, C)]

The Formula is reduced to boundary case

$$F(AC, A) = \text{Best} \begin{cases} F(AC, \_) + F(\_, A) \\ F(A, \_) + S(C, A) \\ F(C, \_) + S(A, A) \end{cases}$$

$$= \text{Best} \begin{cases} -20 - 10 = -30 \\ -10 - 7 = -17 \\ 2 - 10 = -8 \end{cases}$$

The optimal alignment is ACCG AC_G or ACCG A_CG

Table representation of the method:

	Gap	A	C	C	G
Gap	0	-10	-20	-30	-40
A	-10	2 F(A, A)	-8	-18	-28
C	-20	-8	4 F(A, A)+F(C, C)	-6 F(AC, AC)+F(C, _) / F(AC, A)+F(C, C)	-16
G	-30	-18	-6	-3	-4 F(ACC, AC)+F(G,G)

Optimal alignment 1:

ACCG

A_CG

Optimal alignment 2:

ACCG

AC_G

- The arrows preserve the path information
- Arrows in blue highlight the direction that gives the best alignment score
- Trace back the arrows to get the optimal alignment