```
def align(
        seq1: str, seq2: str, match_award=-3, indel_penalty=5, sub_penalty=1, banded_width=-1,
        gap='-'
) -> tuple[float, str | None, str | None]:
   m, n = len(seq1), len(seq2)
   # Initialize the DP table with default value of float('inf')
   dp = \{\}
   dp[(0, 0)] = 0
   for i in range(1, m + 1):
        dp[(i, 0)] = i * indel_penalty
   for j in range(1, n + 1):
        dp[(0, j)] = j * indel_penalty
   if banded_width == -1:
        # Fill in the DP table with full alignment
        for i in range(1, m + 1):
            for j in range(1, n + 1):
                match = dp.get((i - 1, j - 1), float('inf')) + (match_award if seq1[i - 1] ==
seq2[j - 1] else sub_penalty)
                delete = dp.get((i - 1, j), float('inf')) + indel_penalty
                insert = dp.get((i, j - 1), float('inf')) + indel_penalty
                dp[(i, j)] = min(match, insert, delete)
        for i in range(1, m + 1):
            for j in range(max(1, i - banded_width), min(n + 1, i + banded_width + 1)):
                match = dp.get((i - 1, j - 1), float('inf')) + (match_award if seq1[i - 1] ==
seq2[j - 1] else sub_penalty)
                delete = dp.get((i - 1, j), float('inf')) + indel_penalty
                insert = dp.get((i, j - 1), float('inf')) + indel penalty
                dp[(i, j)] = min(match, insert, delete)
   # Backtracking to reconstruct the optimal alignment
   alignment_seq1, alignment_seq2 = [], []
   i, j = m, n
   while i > 0 or j > 0:
        current_cost = dp.get((i, j), float('inf'))
        if i > 0 and j > 0 and (seq1[i - 1] == seq2[j - 1] or current_cost == dp.get((i - 1, j
 1), float('inf')) + sub penalty):
```

```
alignment_seq1.append(seq1[i - 1])
        alignment_seq2.append(seq2[j - 1])
        i -= 1
        j -= 1
    elif i > 0 and current_cost == dp.get((i, j - 1), float('inf')) + indel_penalty:
        alignment_seq1.append(gap)
        alignment_seq2.append(seq2[j - 1])
    else:
        alignment_seq1.append(seq1[i - 1])
        alignment_seq2.append(gap)
        i -= 1
# Reverse the alignments to get the correct order
alignment_seq1.reverse()
alignment_seq2.reverse()
alignment_cost = dp.get((m, n), float('inf'))
return alignment_cost, ''.join(alignment_seq1), ''.join(alignment_seq2)
```

initialization of the DP

```
dp {}
for i in range(1, m + 1): # 0(m)
    dp[(i, 0)] = i * indel_penalty
for j in range(1, n + 1): # 0(n)
    dp[(0, j)] = j * indel_penalty
```

The first row and column are initialized in 0(m + n) times where m = length of seq1 and n = length of seq2.

A dictionary is used to store the dp values. In the case of unrestricted alignment, the dp requires a space complexity of 0(m*n) since it's a 2d structure of size (m+1)*(n+1) and one entry is needed for each cell.

Filling the DP table

```
if banded_width == -1:
    for i in range(1, m + 1): # 0(m)
        for j in range(1, n + 1): # 0(n)
```

the unrestricted alignment (banded_width = -1), the loops iterate through all the cells of the table which leads to 0(m*n)

The rest of the line of code after "for j in range(1, n + 1):" take a constant time of 0(1)

The inner loop of the banded alignment is constrained by the banded width. So for each 'i', the 'j' index stretches for a limited range. If the banded width is k then the complexity of the alignment then becomes 0(m * k).

The band_width is used to constrain the matrix size. As the band for each row is being calculated the space complexity is 0(k*n)

Backtracking

```
while i > 0 or j > 0:
```

the backtracking runs back the alignment by iterating through the sequence which takes 0(m + n) times in both cases.

The operations within the loop is O(1)

<u>Therefore</u>, for **unrestricted alignment**:

Time complexity

Initialization: O(m + n)
Filling the Table: O(m * n)
Backtracking: O(m + n)

■ Total time complexity: O(m * n)

Space Complexity

Space complexity: 0(m*n)

And for **banded alignment:**

Time complexity

Initialization: O(m + n)
Filling the Table: O(m * k)
Backtracking: O(m + n)

■ Total time complexity: O(m * k)

Space complexity

Space complexity: 0(k*n)

Dependency Pointers

In the banded algorithm, the dependency pointers are modified to make sure that the cells within the band are accessible. The algorithm considers cells of (i-1, j-1), (i-1, j), and (I, j-1) if they are within the band. This is to make sure the band doesn't access cells outside of the band.