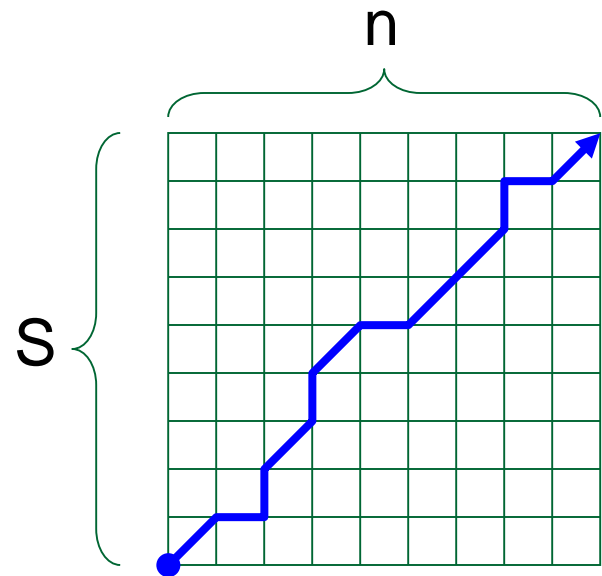


# Computing Knapsack Requires Quadratic Memory

## Solution Path

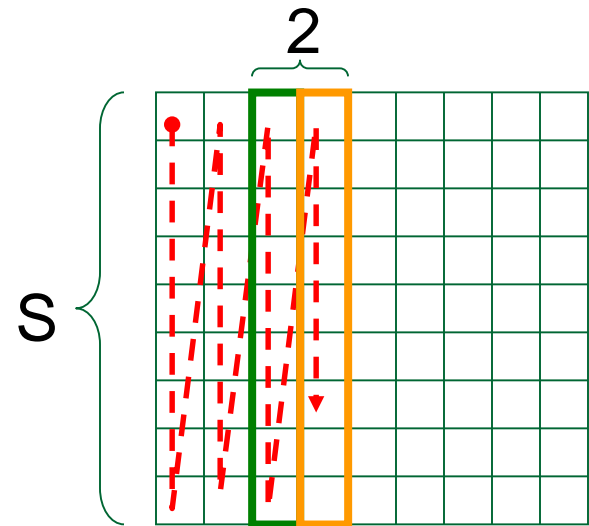
- Space complexity for computing objects used with  $n$  and size  $S$  is  $O(nS)$
- We need to keep all backtracking references in memory to reconstruct the path (backtracking)



# Computing Knapsack Score with Linear Memory

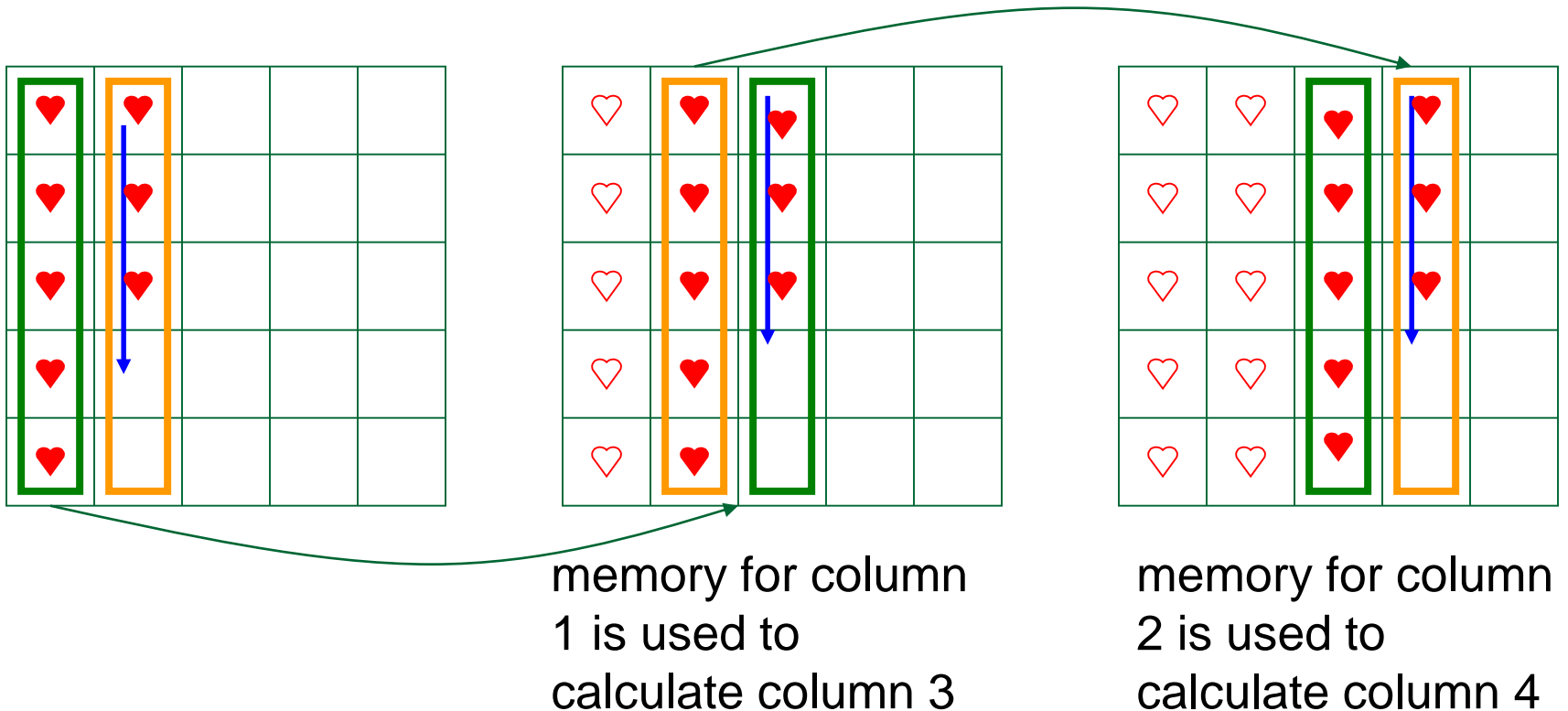
## Knapsack Score

- Space complexity of computing just the score itself is  $O(n)$
- We only need the previous column to calculate the current column, and we can then throw away that previous column once we're done using it

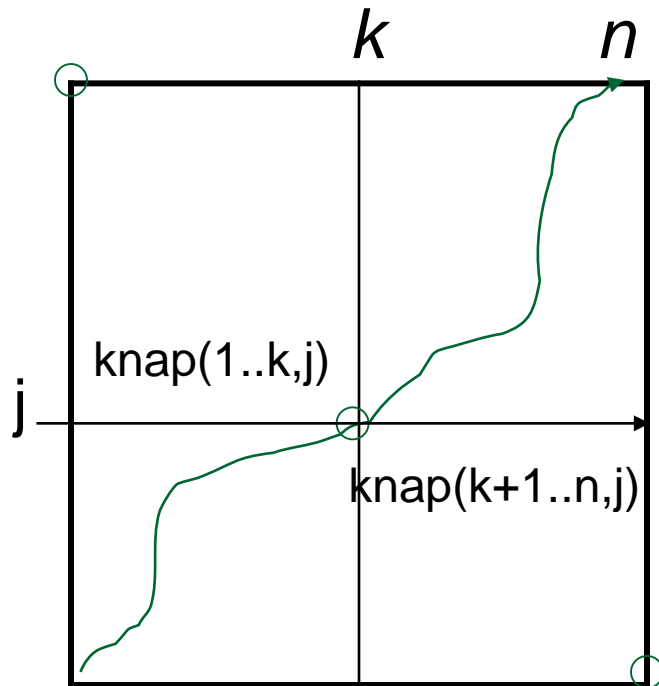


# Computing Knapsack Score: Recycling Columns

Only two columns of scores are saved at any given time



# Crossing the Middle Line



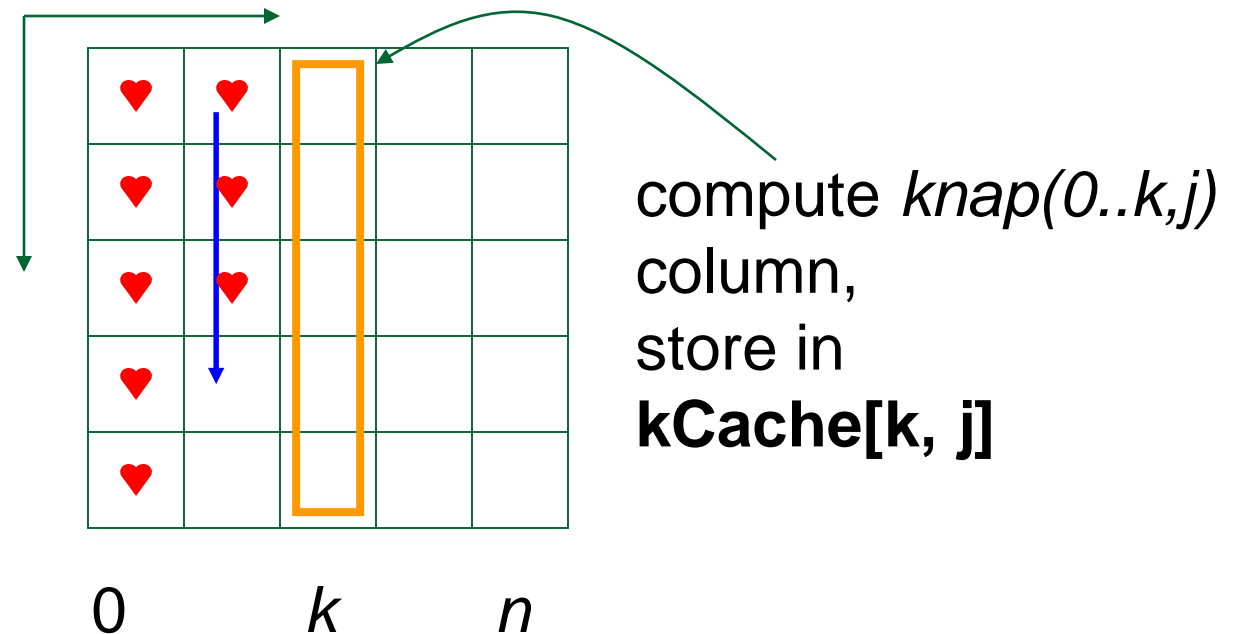
We want to calculate the greatest value paths from  $(0,0)$  to  $(n,S)$  that pass through  $(k,j)$  where  $0 \leq j \leq S$

Define *bestValue*( $j$ )

as the path with the greatest value from  $(0,0)$  to  $(n,S)$  that passes through vertex  $(k,j)$

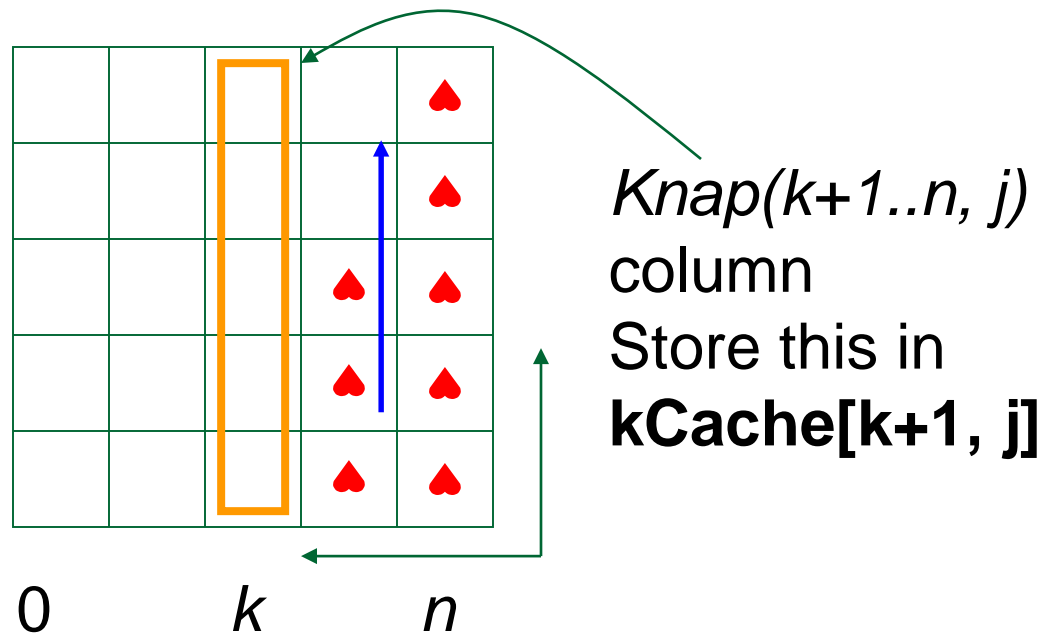
# Computing $\text{knap}(0..k, j)$

- $\text{knap}(0..k, j)$  is the highest value path from  $(0,0)$  to  $(k, j)$  for all  $0 \leq j \leq S$
- Compute  $\text{knap}(0..k, j)$  by dynamic programming in the left half of the matrix



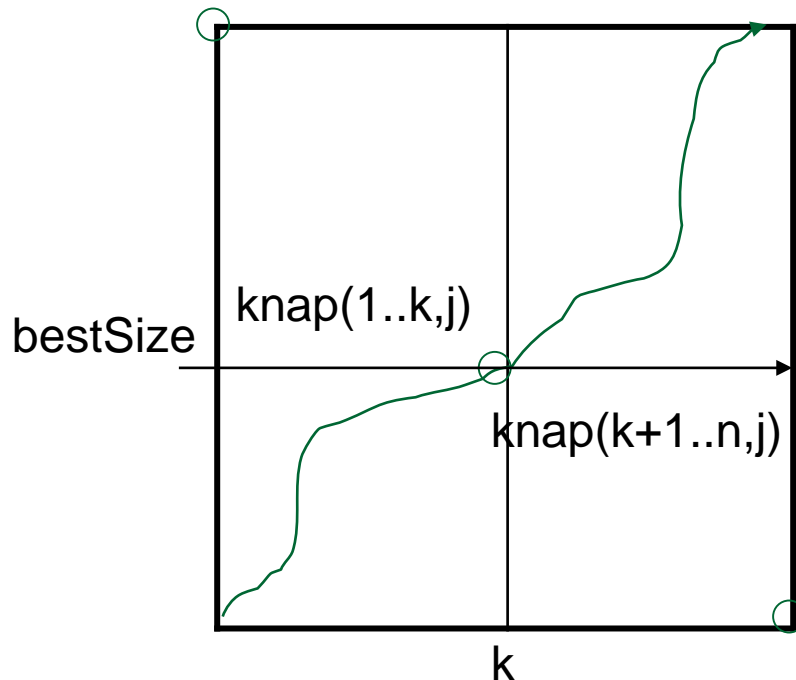
# Computing $\text{knap}(k+1..n, j)$

- $\text{knap}(k+1..n, j)$  is the highest value path from  $(n, S)$  to  $(k+1, j)$  to for all  $j$ ,  $1 \leq j \leq S$
- Compute  $\text{knap}(k+1..n, j)$  by dynamic programming in the right half of the “reversed” matrix



$$bestValue(j) = knap(0..k,j) + knap(k+1..n,j)$$

- For each  $j$  add  $knap(0..k,j)$  to  $knap(k+1..n,j)$  to get the  $bestValue(j)$
- Find the  $j$  value where  $bestValue(j)$  is maximized to find the crossing point
- $bestValue = \text{Max}(bestValue(j))$
- $bestSize = \text{argMax}(bestValue(j)), \quad 0 \leq j \leq S$

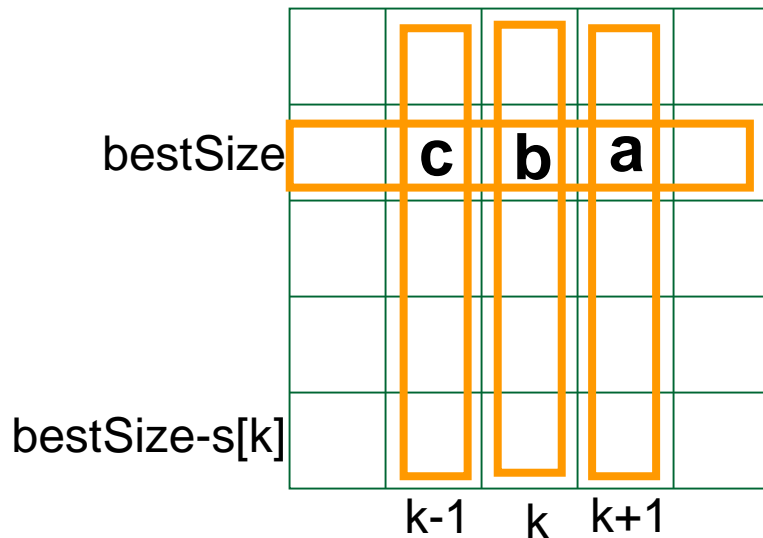


We have found the coordinate of the crossing point of the global solution at index  $k$ :  $(k, bestSize)$

Note this will not necessarily be the middle of  $S$

## *Now we determine if object $k$ is used for the best solution*

- We have bestValue, bestSize and the results of the two DP scans
- Consider a close up of the cached results



**a**  $k\text{Cache}[k+1, \text{bestSize}]$

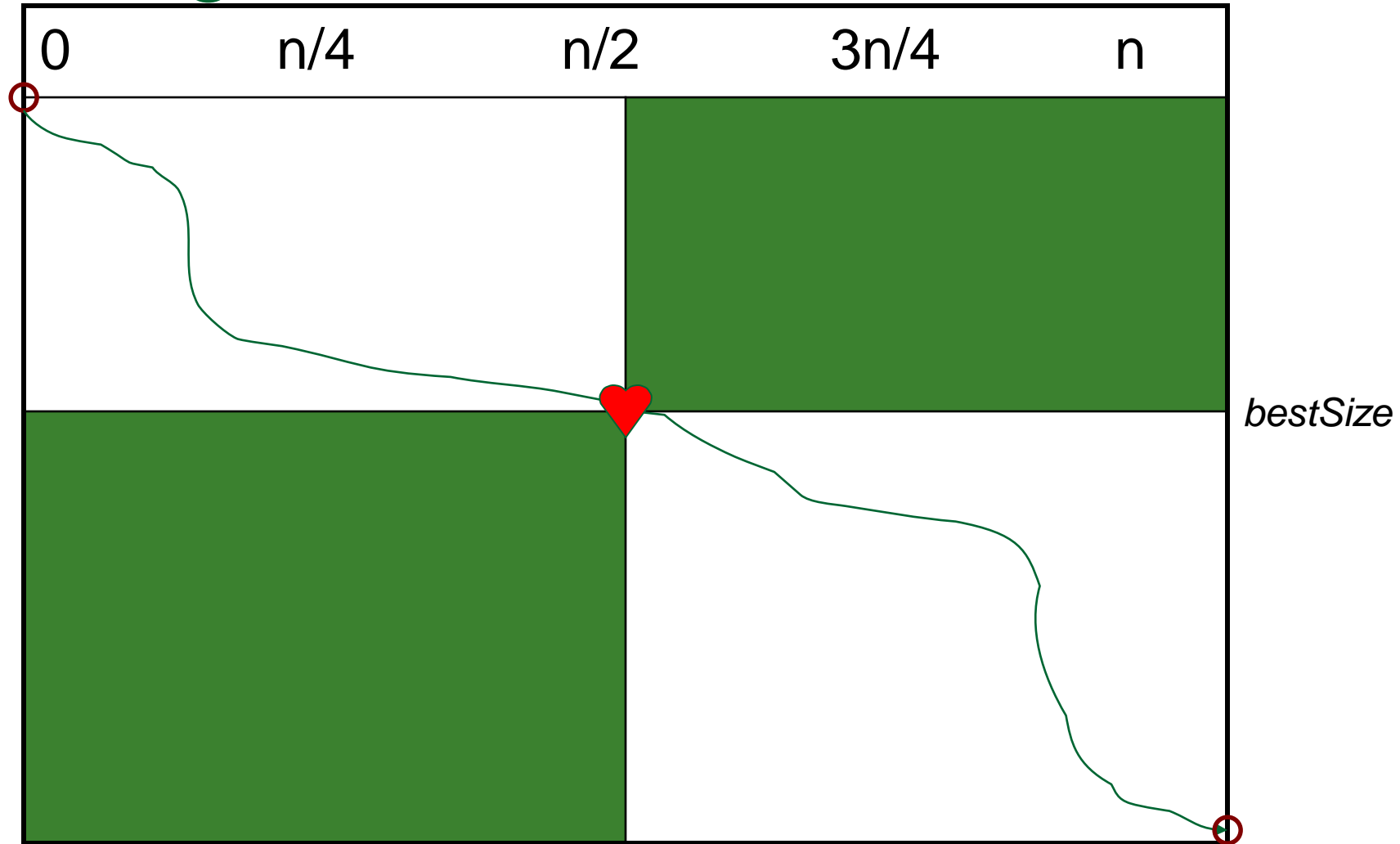
**b**  $k\text{Cache}[k, \text{bestSize}]$

**c**  $k\text{Cache}[k-1, \text{bestSize}]$

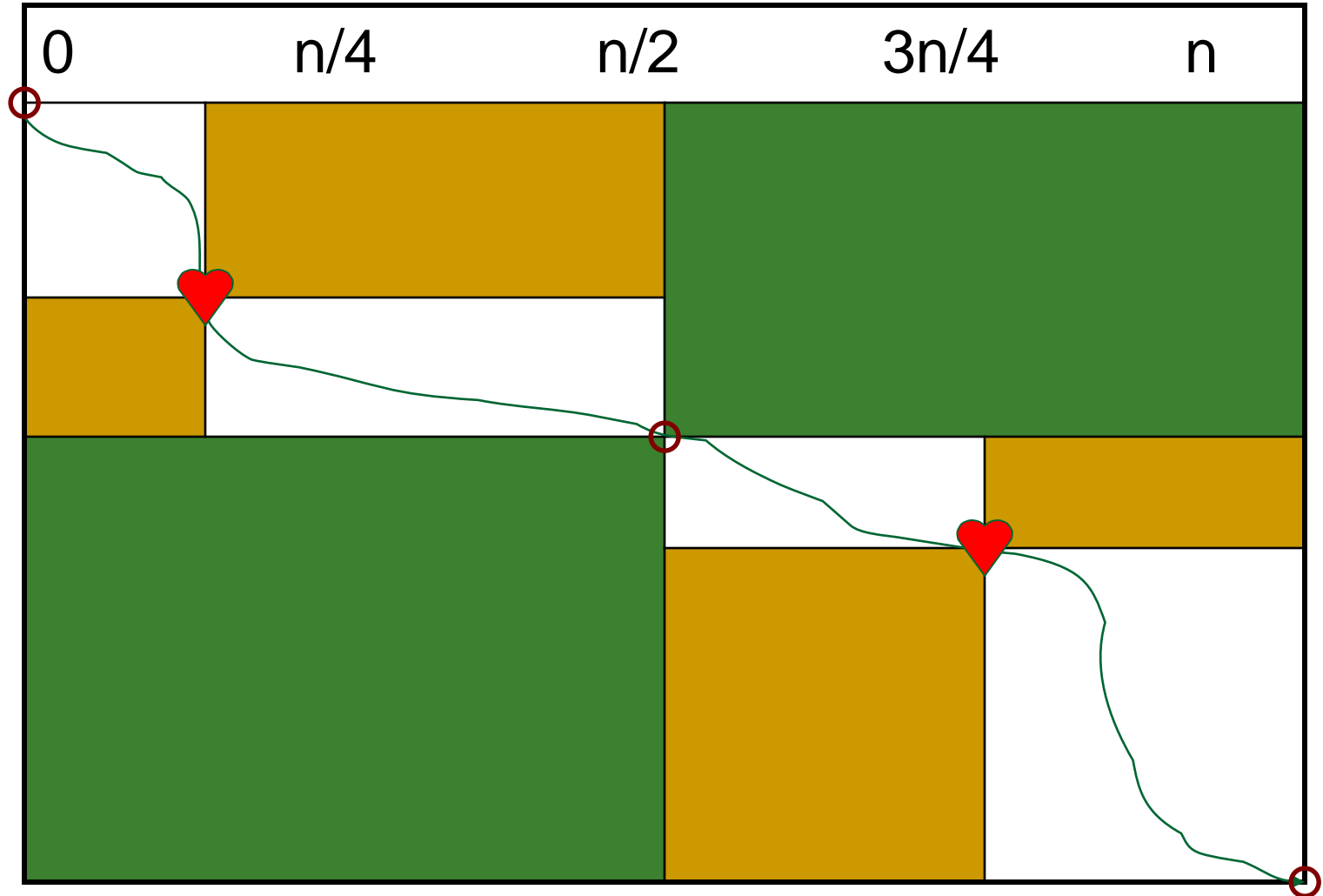
Use object  $k$  is true when (**b**!=**c**) is true



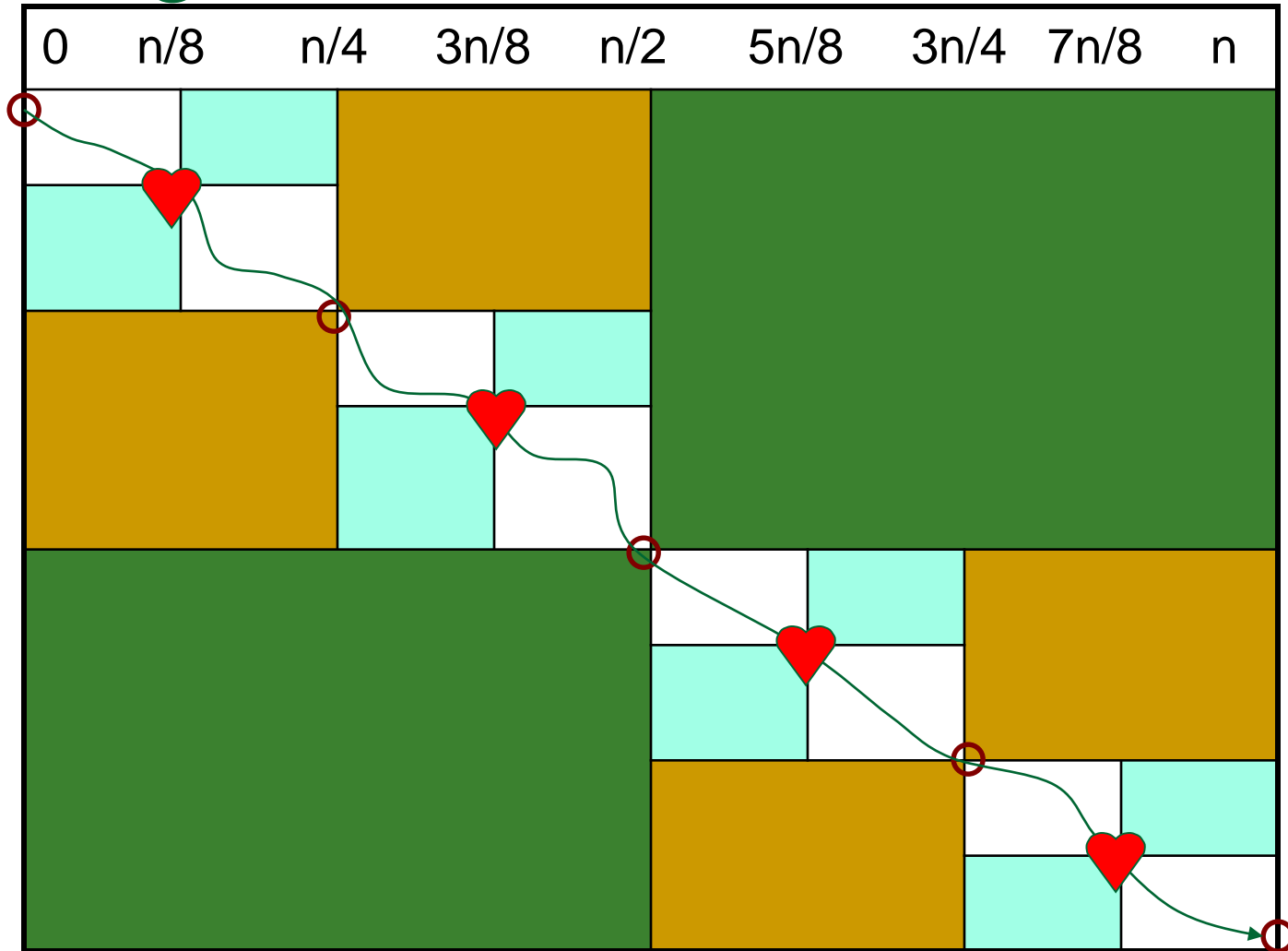
# Finding the Middle Point



# Finding the Middle Point again



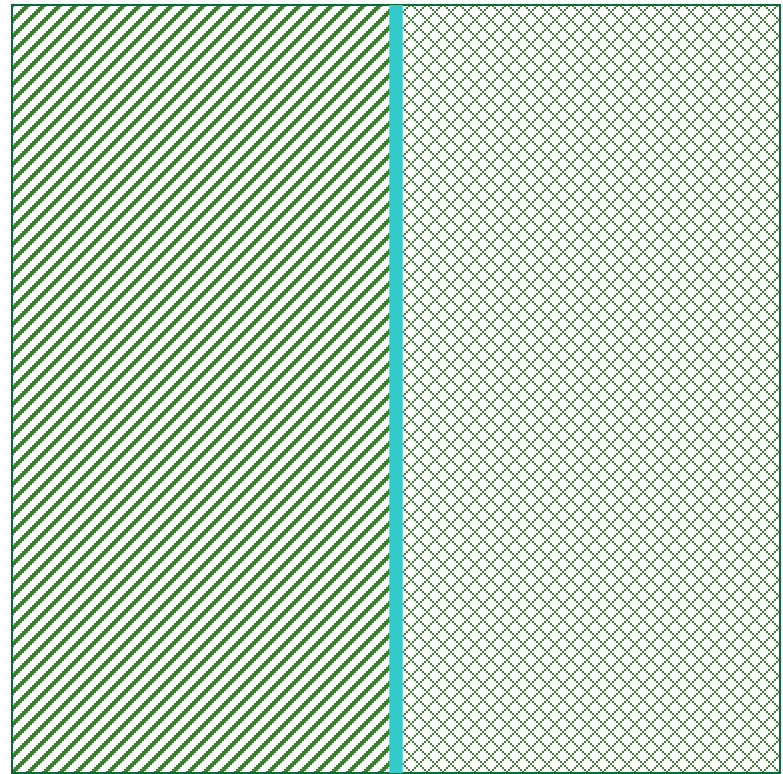
# And Again



# Time = Area: First Pass

- On first pass, the algorithm covers the entire area

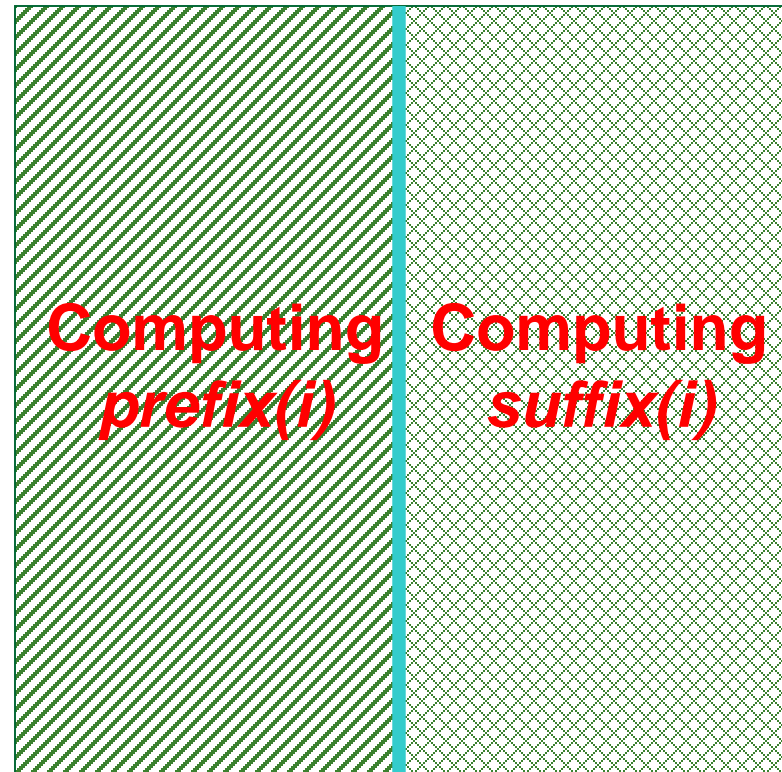
$$\text{Area} = n \bullet m$$



# Time = Area: First Pass

- On first pass, the algorithm covers the entire area

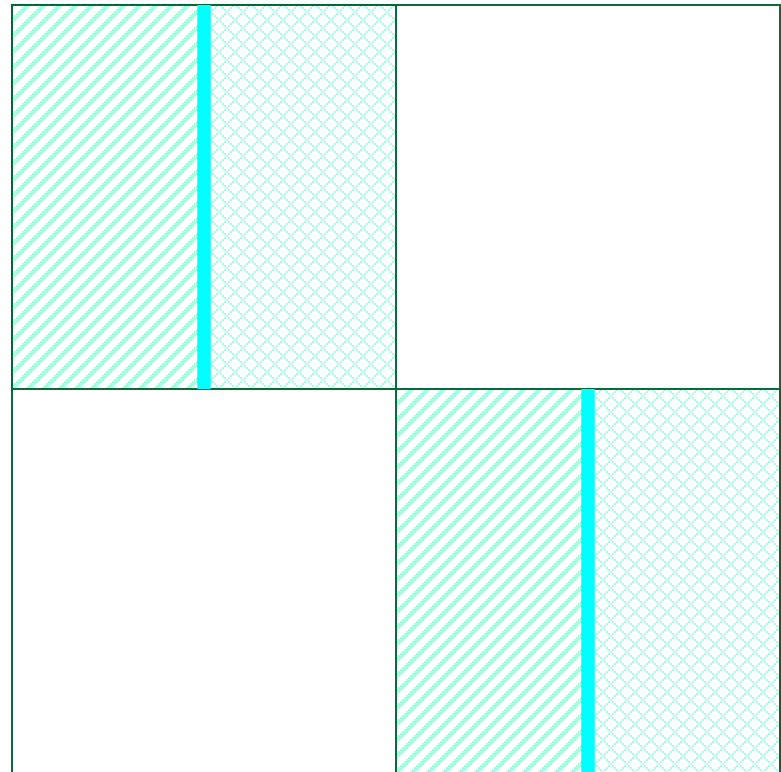
$$\text{Area} = n \bullet m$$



# Time = Area: Second Pass

- On second pass, the algorithm covers only  $1/2$  of the area

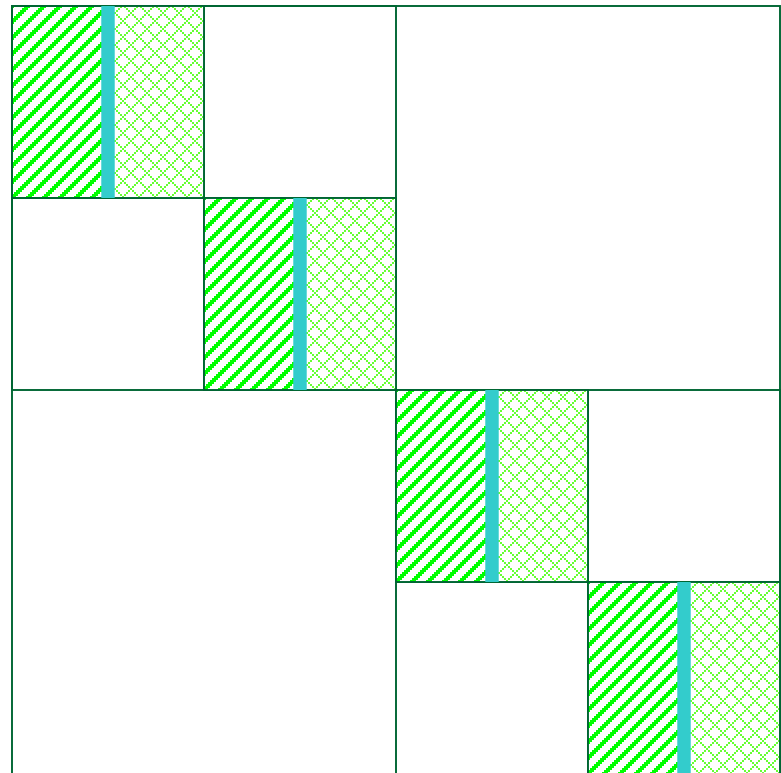
Area/2



# Time = Area: Third Pass

- On third pass, only 1/4th is covered.

Area/4



# Geometric Reduction At Each Iteration

$$1 + \frac{1}{2} + \frac{1}{4} + \dots + (\frac{1}{2})^k \leq 2$$

- Runtime:  $O(\mathbf{Area}) = O(nm)$

