

CSE 421 Algorithms

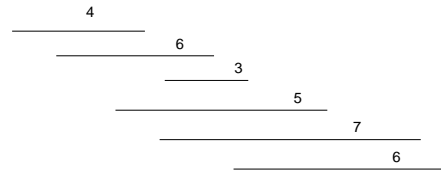
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Lecture 16, Winter 2019
Dynamic Programming

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Intervals sorted by end time

Dynamic Programming

- Weighted Interval Scheduling
- Given a collection of intervals I_1, \dots, I_n with weights w_1, \dots, w_n , choose a maximum weight set of non-overlapping intervals



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Intervals sorted by end time

Optimality Condition

- $Opt[j]$ is the maximum weight independent set of intervals I_1, I_2, \dots, I_j
- $Opt[j] = \max(Opt[j-1], w_j + Opt[p[j]])$
 - Where $p[j]$ is the index of the last interval which finishes before I_j starts

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Algorithm

```
MaxValue(j) =  
  if j = 0 return 0  
  else  
    return max( MaxValue(j-1),  
               w_j + MaxValue(p[j]) )
```

Worst case run time: 2^n

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A better algorithm

$M[j]$ initialized to -1 before the first recursive call for all j

```
MaxValue(j) =  
  if j = 0 return 0;  
  else if  $M[j] \neq -1$  return  $M[j]$ ;  
  else  
     $M[j] = \max(\text{MaxValue}(j-1), w_j + \text{MaxValue}(p[j]))$ ;  
    return  $M[j]$ ;
```

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Iterative Algorithm

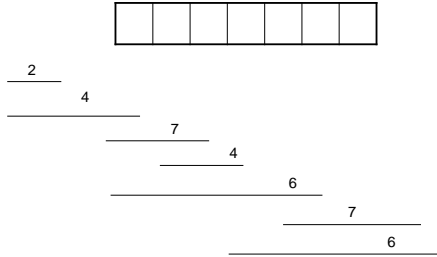
Express the MaxValue algorithm as an iterative algorithm

```
MaxValue {  
  
  
  
  
  
  
}
```

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Fill in the array with the Opt values

$$\text{Opt}[j] = \max(\text{Opt}[j - 1], w_j + \text{Opt}[p[j]])$$

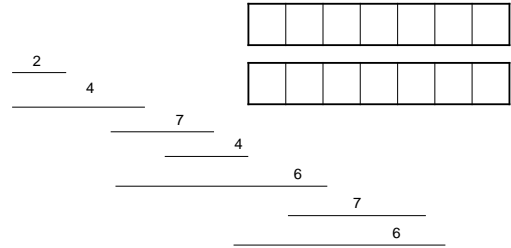


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Computing the solution

$$\text{Opt}[j] = \max(\text{Opt}[j - 1], w_j + \text{Opt}[p[j]])$$

Record which case is used in Opt computation



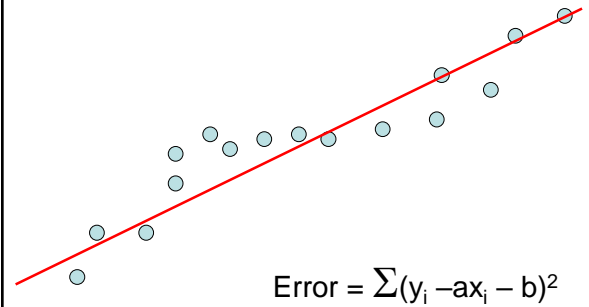
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Dynamic Programming

- The most important algorithmic technique covered in CSE 421
- Key ideas
 - Express solution in terms of a polynomial number of sub problems
 - Order sub problems to avoid recomputation

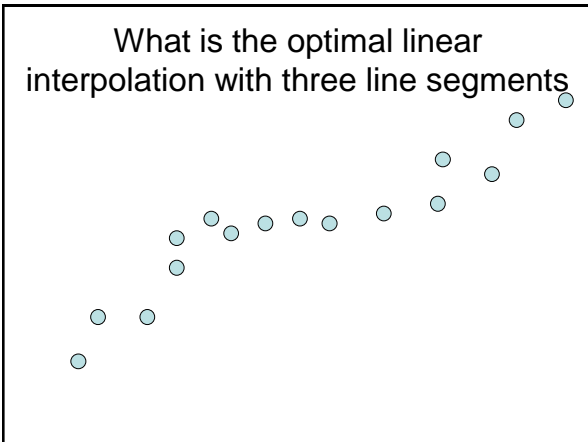
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Optimal linear interpolation



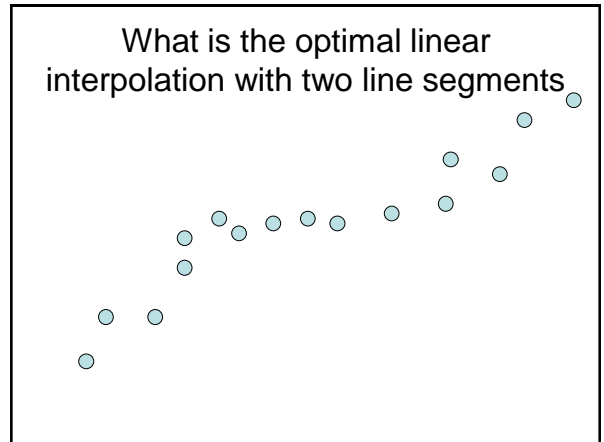
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What is the optimal linear interpolation with three line segments



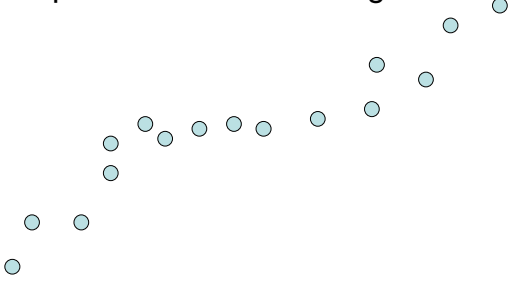
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What is the optimal linear interpolation with two line segments



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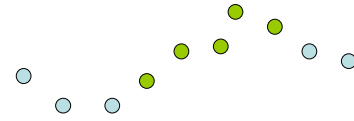
What is the optimal linear interpolation with n line segments



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Notation

- Points p_1, p_2, \dots, p_n ordered by x-coordinate ($p_i = (x_i, y_i)$)
- $E_{i,j}$ is the least squares error for the optimal line interpolating p_i, \dots, p_j



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Optimal interpolation with two segments

- Give an equation for the optimal interpolation of p_1, \dots, p_n with two line segments
- $E_{i,j}$ is the least squares error for the optimal line interpolating p_i, \dots, p_j

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Optimal interpolation with k segments

- Optimal segmentation with three segments
 - $\text{Min}_{i,j} \{E_{1,i} + E_{i,j} + E_{j,n}\}$
 - $O(n^2)$ combinations considered
- Generalization to k segments leads to considering $O(n^{k-1})$ combinations

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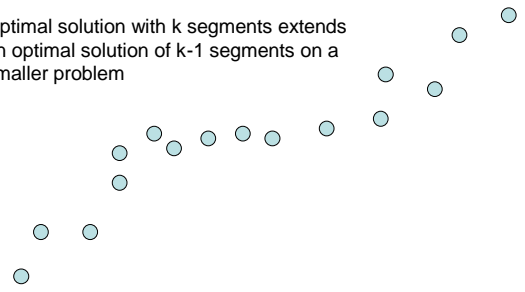
$\text{Opt}_k[j]$: Minimum error approximating $p_1 \dots p_j$ with k segments

How do you express $\text{Opt}_k[j]$ in terms of $\text{Opt}_{k-1}[1], \dots, \text{Opt}_{k-1}[j]$?

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Optimal sub-solution property

Optimal solution with k segments extends an optimal solution of k-1 segments on a smaller problem



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Optimal multi-segment interpolation

Compute $\text{Opt}[k, j]$ for $0 < k < j < n$

```
for j := 1 to n
  Opt[ 1, j ] =  $E_{1,j}$ ;
for k := 2 to n-1
  for j := 2 to n
    t :=  $E_{1,j}$ 
    for i := 1 to j -1
      t = min (t,  $\text{Opt}[k-1, i ] + E_{i,j}$ )
    Opt[k, j] = t
```

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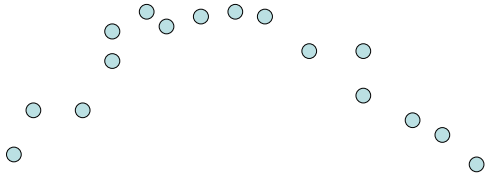
Determining the solution

- When $\text{Opt}[k,j]$ is computed, record the value of i that minimized the sum
- Store this value in a auxiliary array
- Use to reconstruct solution

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Variable number of segments

- Segments not specified in advance
- Penalty function associated with segments
- Cost = Interpolation error + $C \times \text{\#Segments}$



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Penalty cost measure

- $\text{Opt}[j] = \min(E_{1,j}, \min_i(\text{Opt}[i] + E_{i,j} + P))$

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