

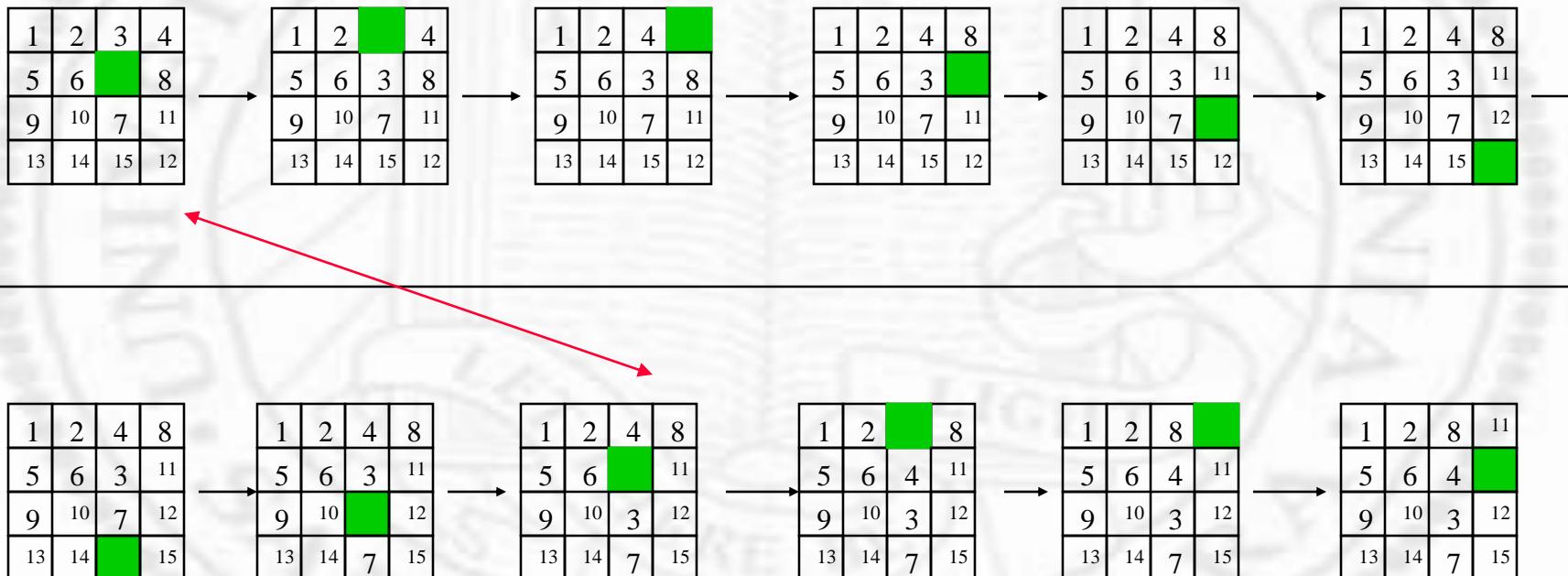
# Branch and Bound

- The 15 puzzle
  - Input: 15 numbered tiles (1-15) and a blank one on a 4x4 square
  - Legal moves: a numbered title, which is a 4-neighbor of the blank title, can move into the location of the blank tile
  - Output: an ordered final configuration



## □ Depth first search

- up, right, down, left order
- do not undo the last move



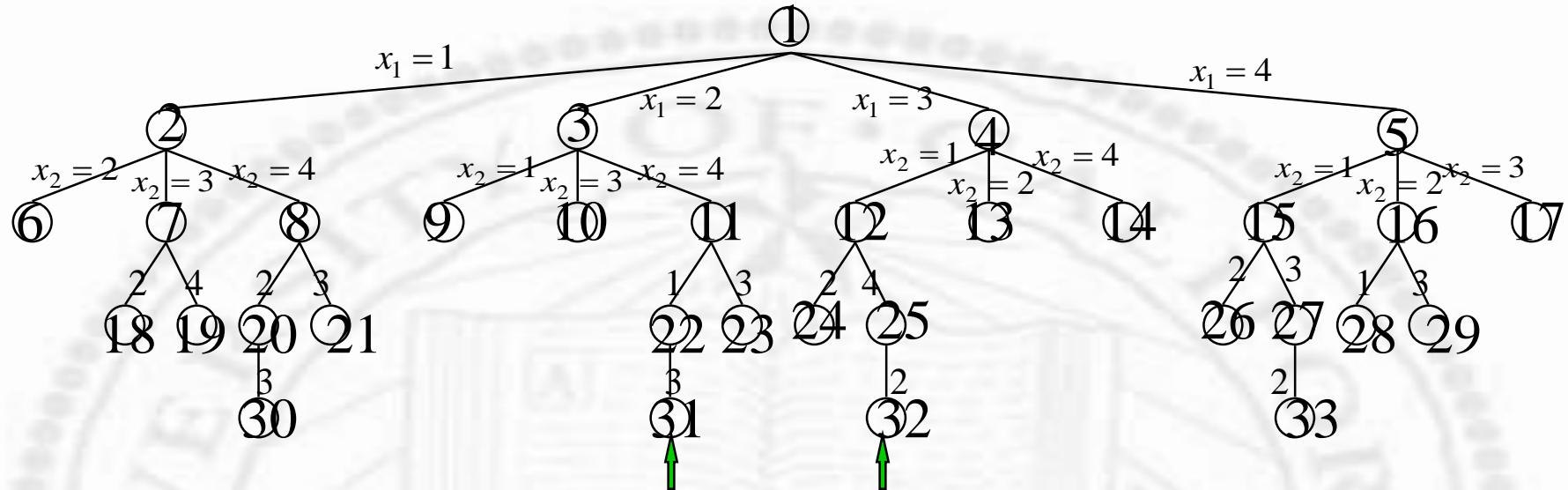
- Depth first search
  - Always go left as deep as possible
  - The left path
    - ◆ may be very long for large search problems
    - ◆ may contain a cycle and never terminate
    - ◆ the optimal solution may not be found there
    - ◆ may not be the natural way of searching
      - e.g. playing chess (many alternatives are considered simultaneously with a small number of look-ahead)



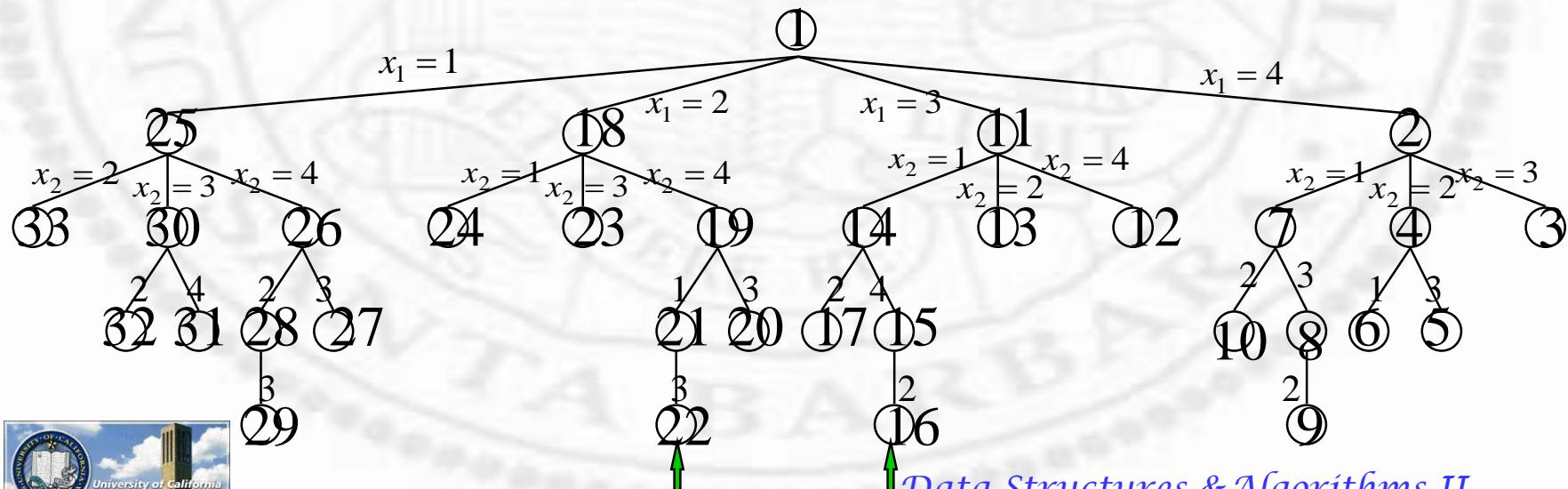
- Breadth search
  - An E-node stays an E-node until all its children are generated
  - How to organize all the children?
    - ♦ FIFO (breadth first)
    - ♦ FILO (D-search)

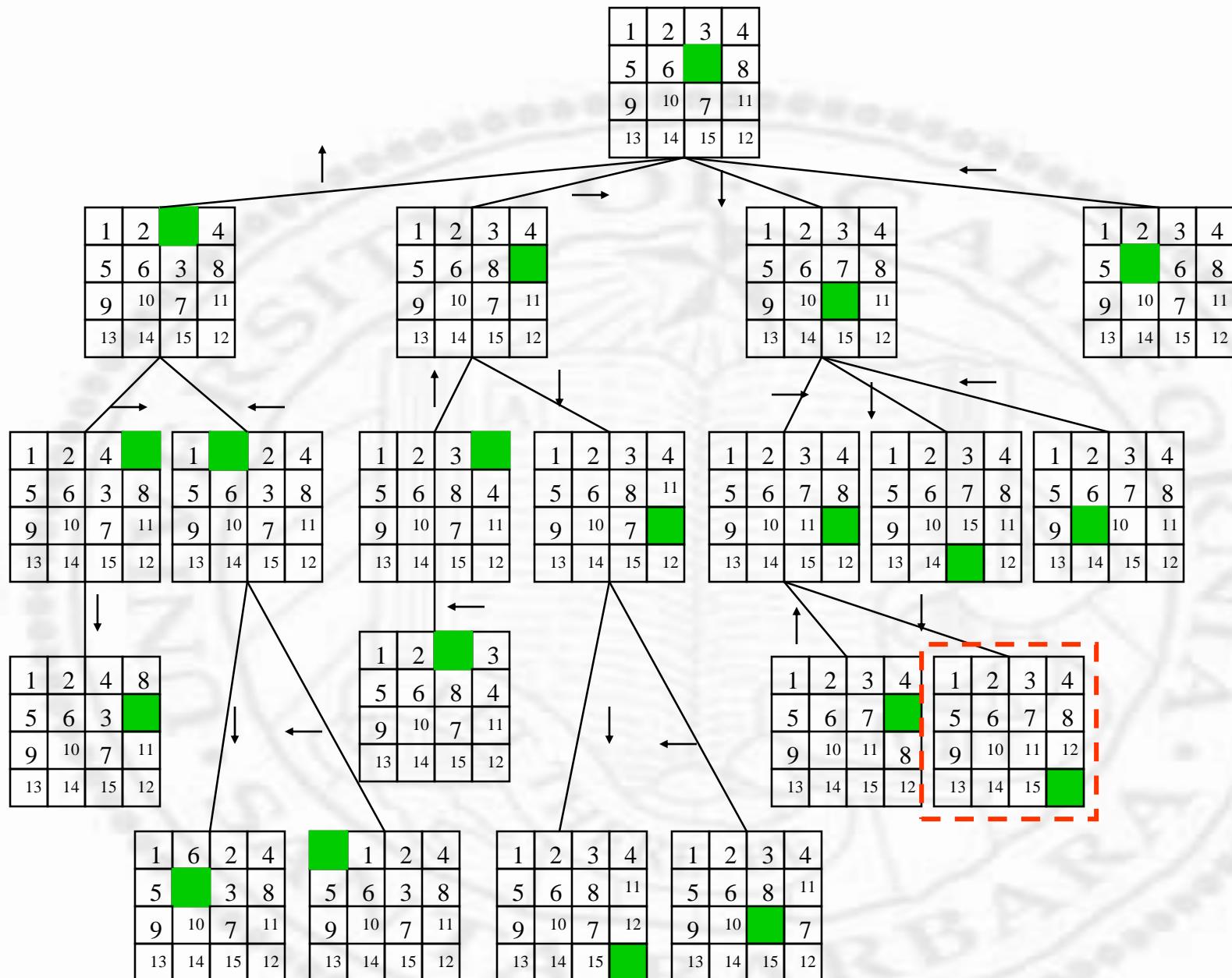


# FIFO breadth first search for 4-queen



# FILO breadth first search for 4-queen

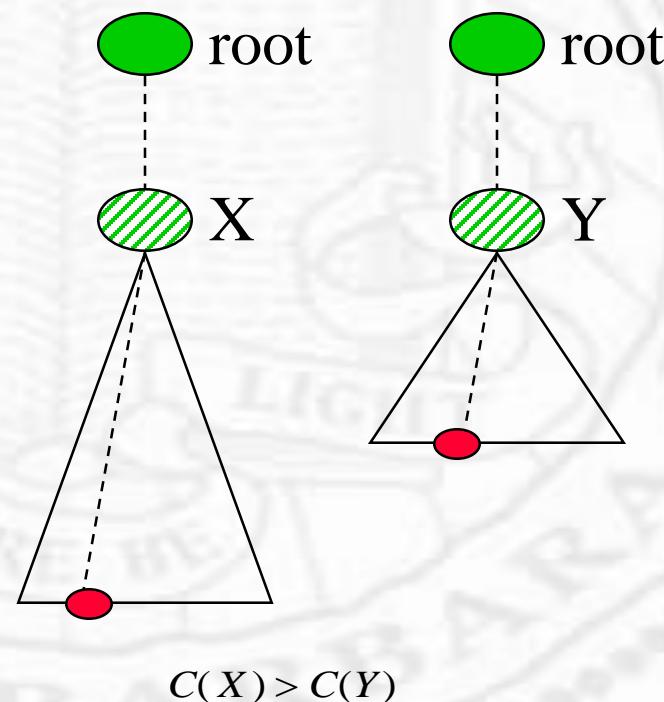
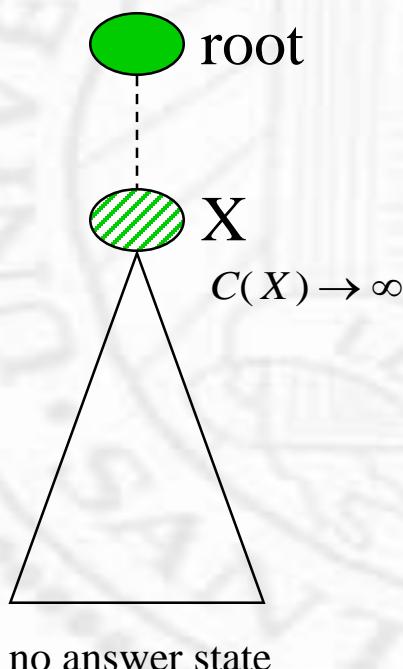




- Problems with depth, breadth searches
  - They have *rigid rules*
- If during the search it becomes apparent that a solution is near, search pattern should be altered to the path(s) that lead quickly to that solution



- Q: How to know which path is “better” and should be explored first?
- A: A cost function that is based on the cost of the solution states reachable



- Q: How do we know the cost of a partial problem state  $X$ ?
- A: Trace out the subtree rooted at  $X$
- But then the subtree has already been explored, no use of the cost function
  - ⇒ Ideal cost function is useless, need an *estimate* without generating the subtree



- How to generate an *educated* guess?
  - current position: cost of reaching current node from root (*known*)
  - estimated distance to an answer state: cost of reaching a solution (*unknown*)

$$\hat{C}(X) = f(h(X)) + \hat{g}(X)$$

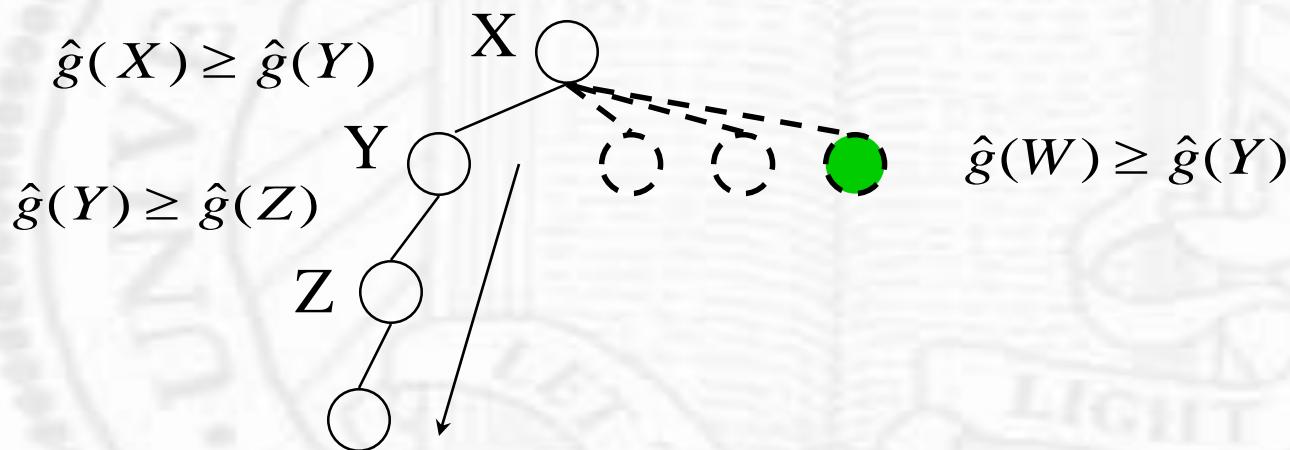
$h(X)$ : distance from root to  $X$

$f(\cdot)$ : nondecreasing function

$\hat{g}(X)$ : *estimated* cost to reach an answer from  $X$



- Use only  $\hat{g}(X)$ 
  - normally, we expect search to make progress, i.e., exploring deeper into the state space tree leads us closer to an answer



- may turn into a depth first search



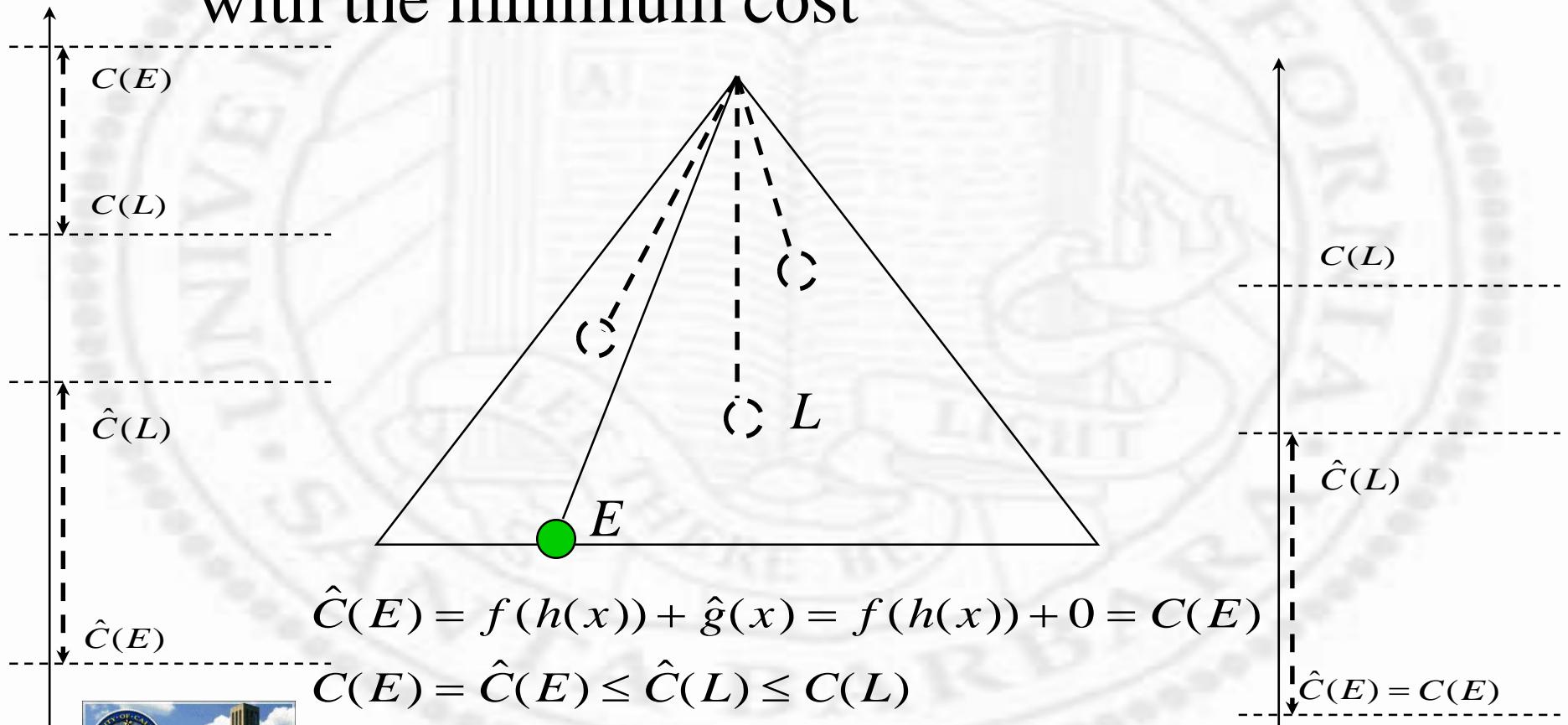
- Use only  $f(h(x))$ 
  - level ↓
  - distance to root ↓
  - effort to get there  $h(x)$  ↓
  - cost  $f(h(x))$  ↓
  - expand earlier
- ⇒ breadth first search
- A *balanced* cost function should have both components



- A search based on the least-cost-first principle is called *LC branch-and-bound* search
  - Can be more efficient
  - Will it find the optimal solution?



- **Proposition:** If the estimated cost function never over-estimates the cost, then LC branch-and-bound find the answer state with the minimum cost



# Trade-off in Search Pattern

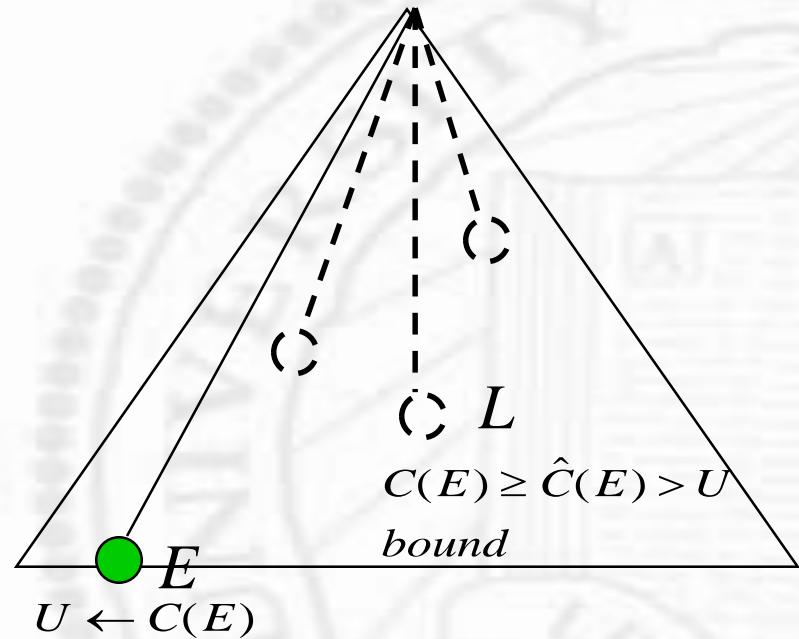
- Depth, breadth, etc.
  - rigid ordering
  - applicable to all
  - no need for a cost function
- LC branch-and-bound
  - more flexible
  - applicable if a good cost function is available

# Trade-off in Bounding

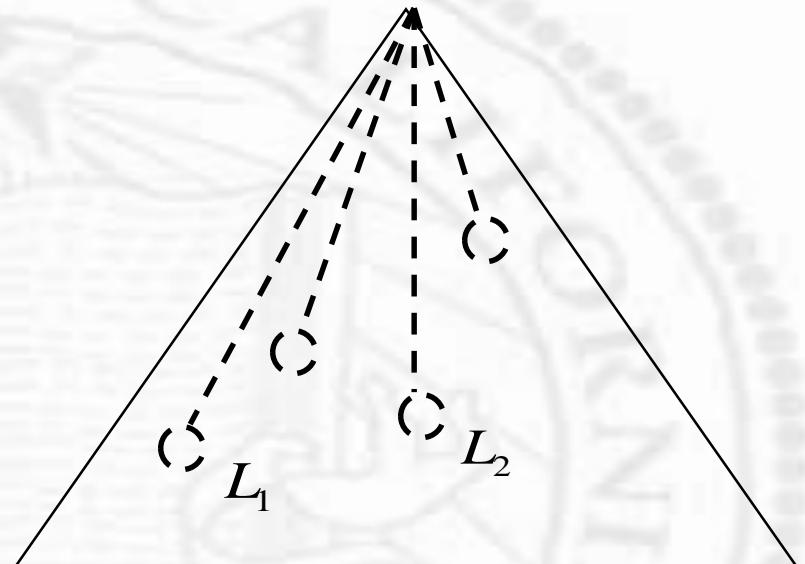
- Depth first
  - feasibility
  - optimality
- Breadth and LC branch-and-bound
  - feasibility
  - optimality (?)



- In back tracking



- In breadth and LC branch and bound



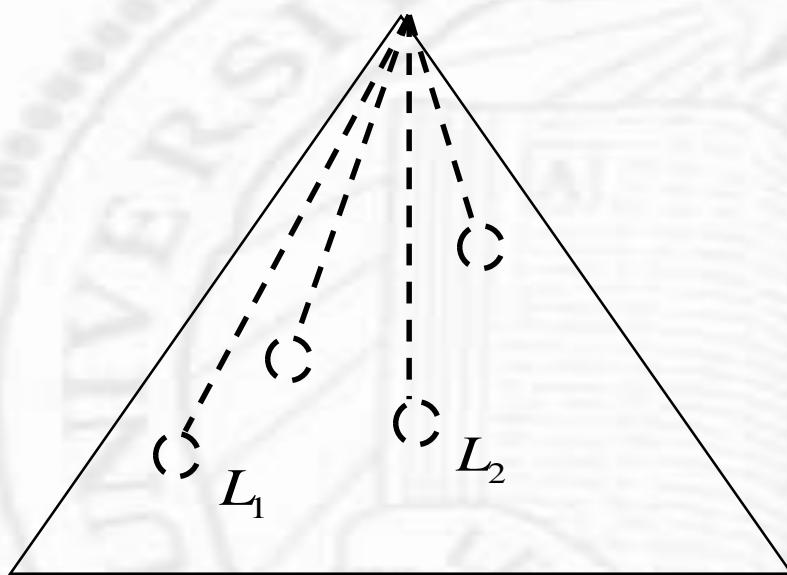
$$\frac{\hat{C}(L_2) \quad \hat{C}(L_1) \quad C(L_2) \quad C(L_1)}{\text{cost}}$$

$$\frac{\hat{C}(L_2) \quad C(L_2) \quad \hat{C}(L_1) \quad C(L_1)}{\text{cost}}$$

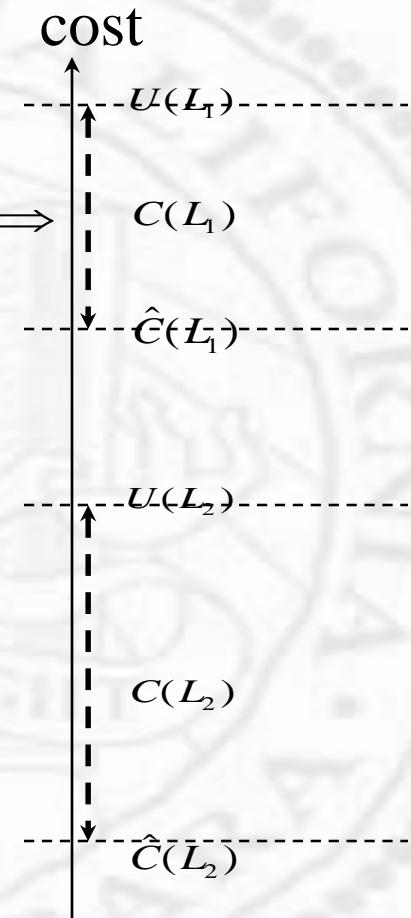
$$\frac{\hat{C}(L_2) \quad \hat{C}(L_1) \quad C(L_1) \quad C(L_2)}{\text{cost}}$$

*Data Structures & Algorithms II*

# Bounding in Branch-and-Bound



$L_1$  is bound  $\Rightarrow$

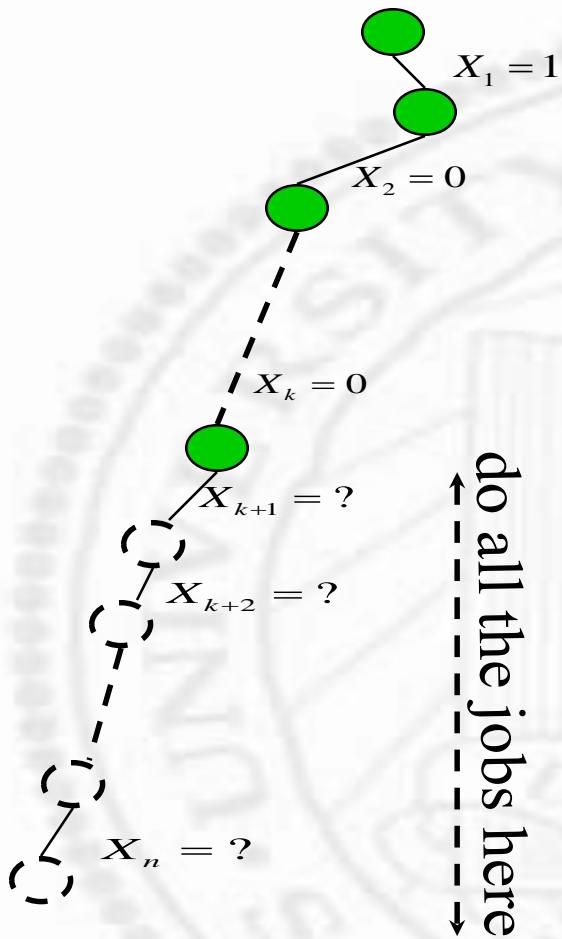


# Job Sequencing with Deadline

- Input:
  - a set of  $n$  jobs with deadlines and penalty
  - a single machine to execute all jobs
- Output: A subset of jobs, each completed before the deadline with least penalty

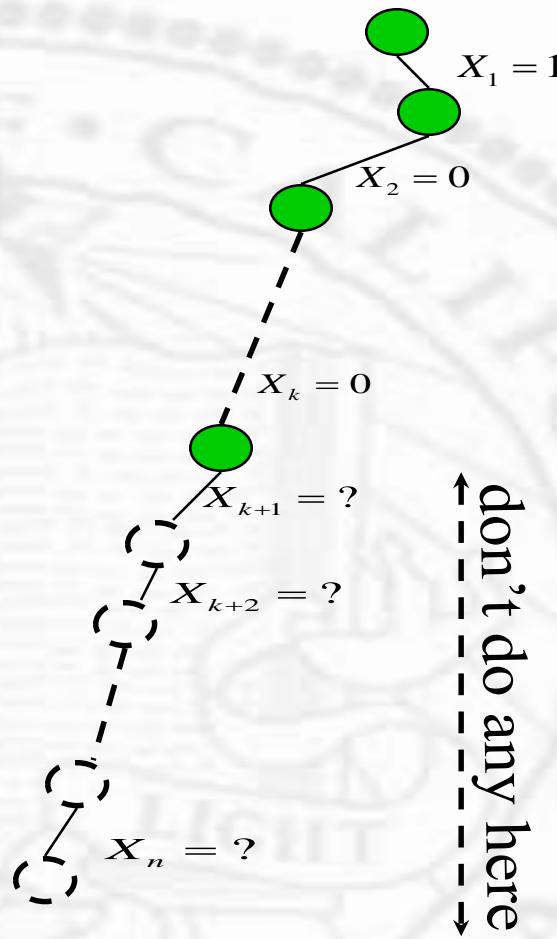


## □ Lower bound



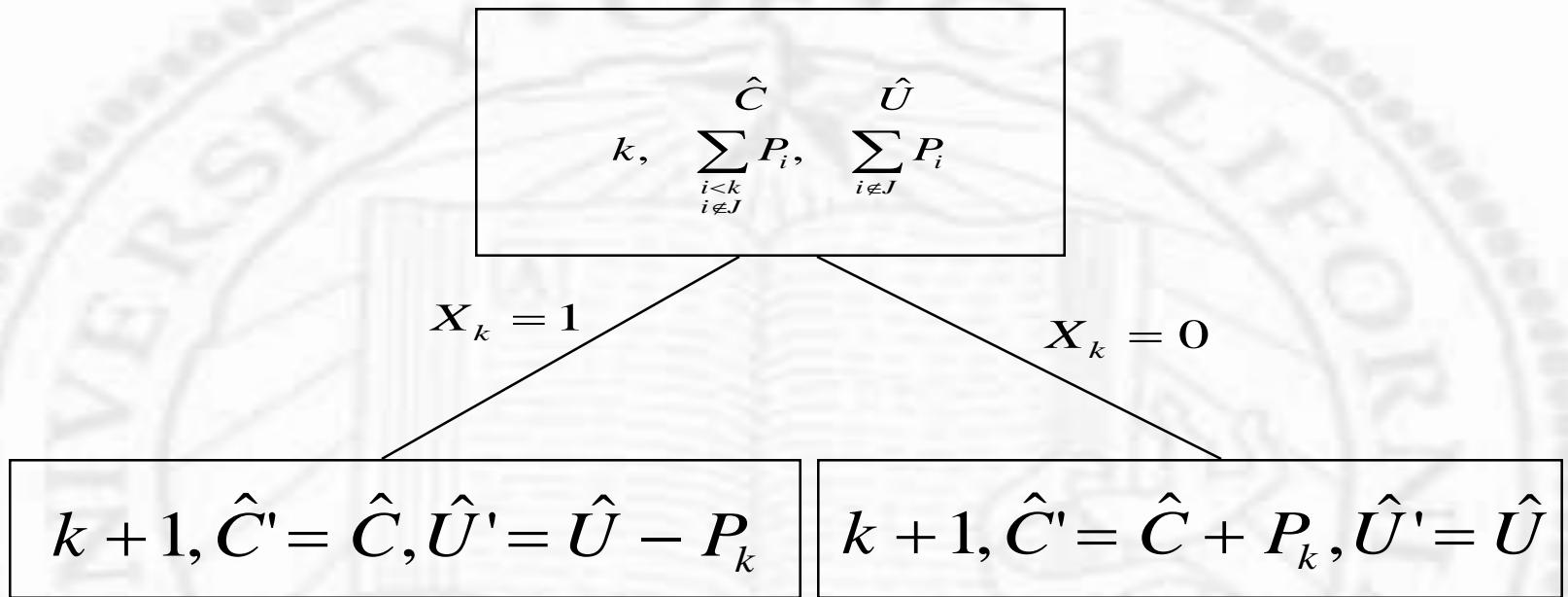
$$\sum_{1 \leq i \leq k} P_i(1 - X_i) = \sum_{\substack{i \leq k \\ i \notin J}} P_i$$

## □ Upper bound



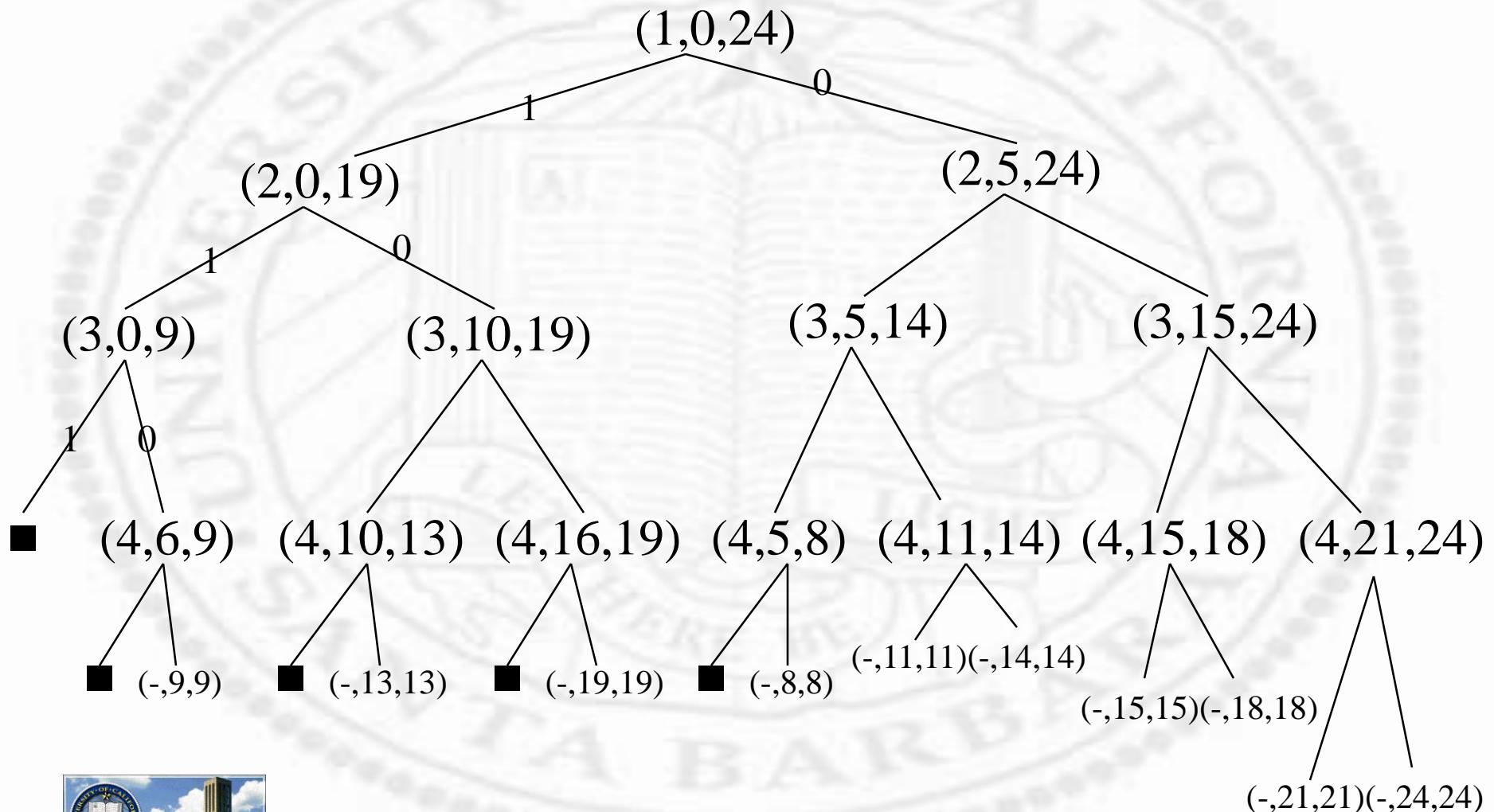
$$\sum_{1 \leq i \leq k} P_i(1 - X_i) + \sum_{i > k} P_i = \sum_{\substack{i \leq k \\ i \notin J}} P_i + \sum_{i > k} P_i$$





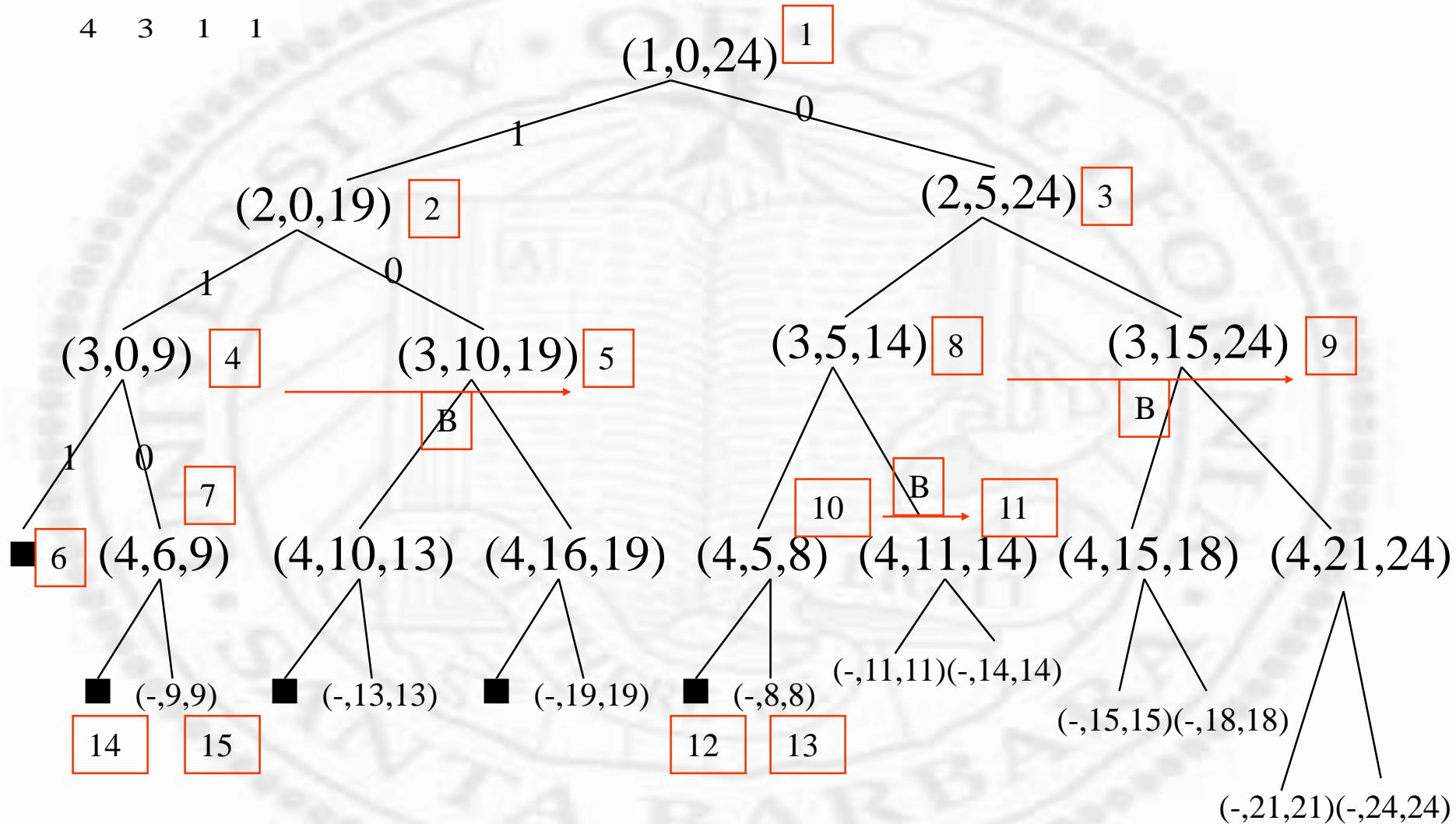
	$p$	$d$	$t$
1	5	1	1
2	10	3	2
3	6	2	1
4	3	1	1

$$k, \quad \sum_{\substack{i < k \\ i \notin J}} P_i, \quad \sum_{i \in J} P_i$$



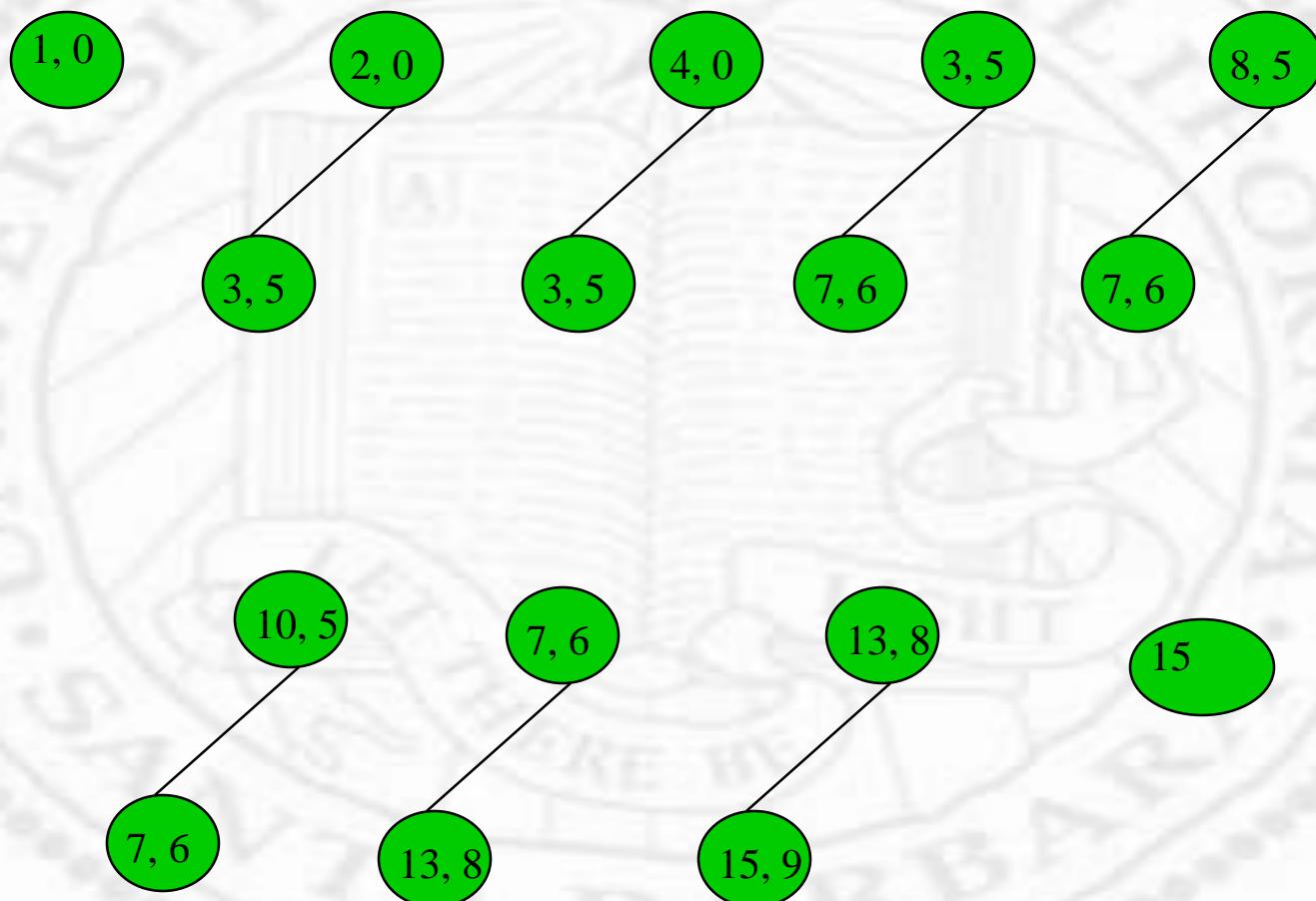
	$p$	$d$	$t$
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$$k, \quad \sum_{\substack{i < k \\ i \notin J}} P_i, \quad \sum_{i \in J} P_i$$

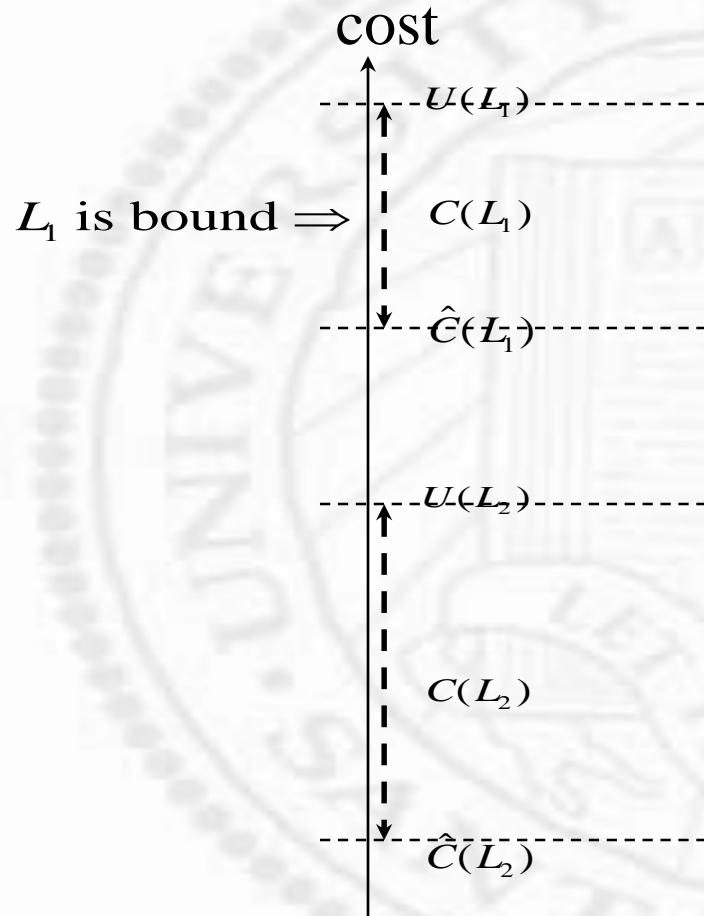


- Store candidates in a partially-ordered tree (a heap)

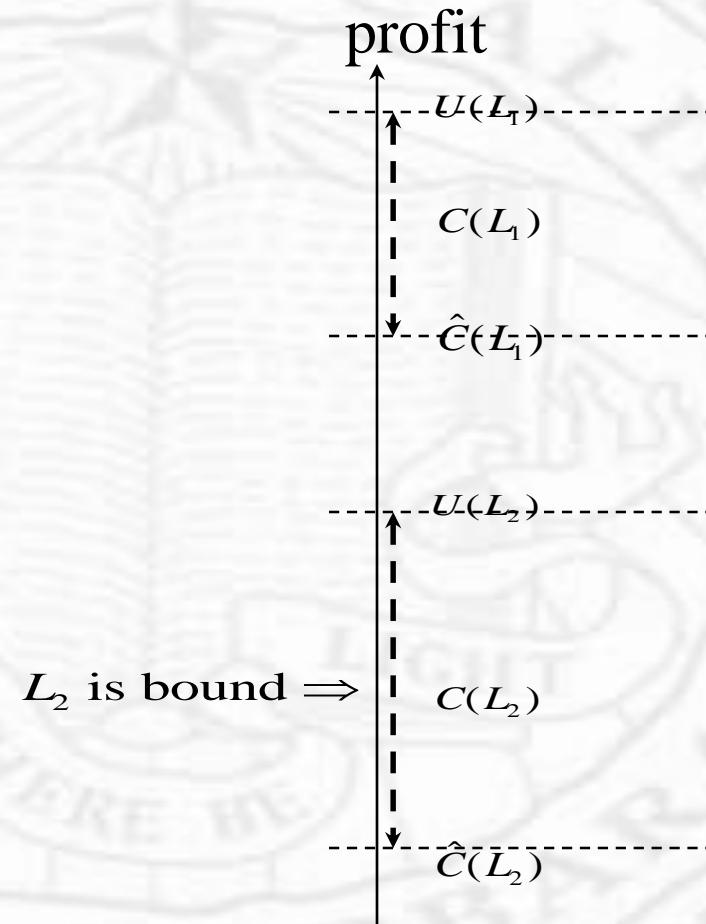
(node number,  $c$ )



- Minimization problems

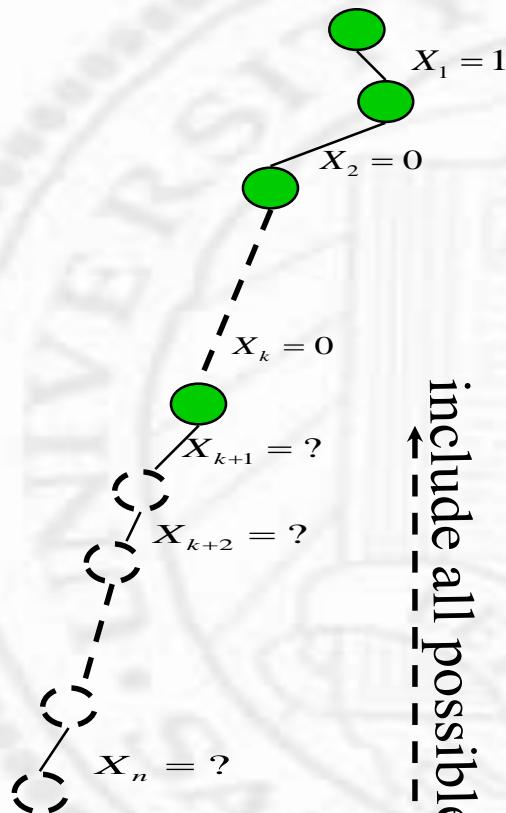


- Maximization problems

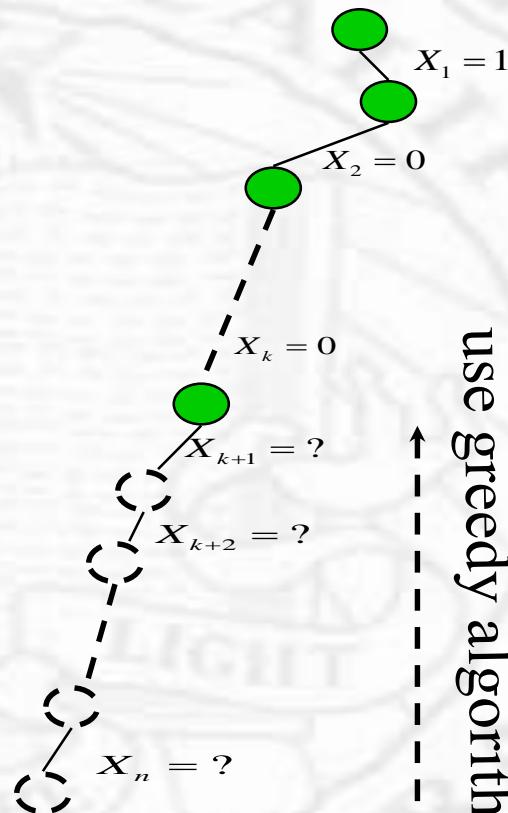


# 0/1-Knapsack

□ Lower bound



□ Upper bound



use greedy algorithm

include all possible

$$k, \sum_{\substack{i < k \\ i \in \text{sack}}} P_i + \text{integer}, \quad \sum_{\substack{i < k \\ i \in \text{sack}}} P_i + \text{greedy}$$

$$X_k = 1$$

$$X_k = 0$$

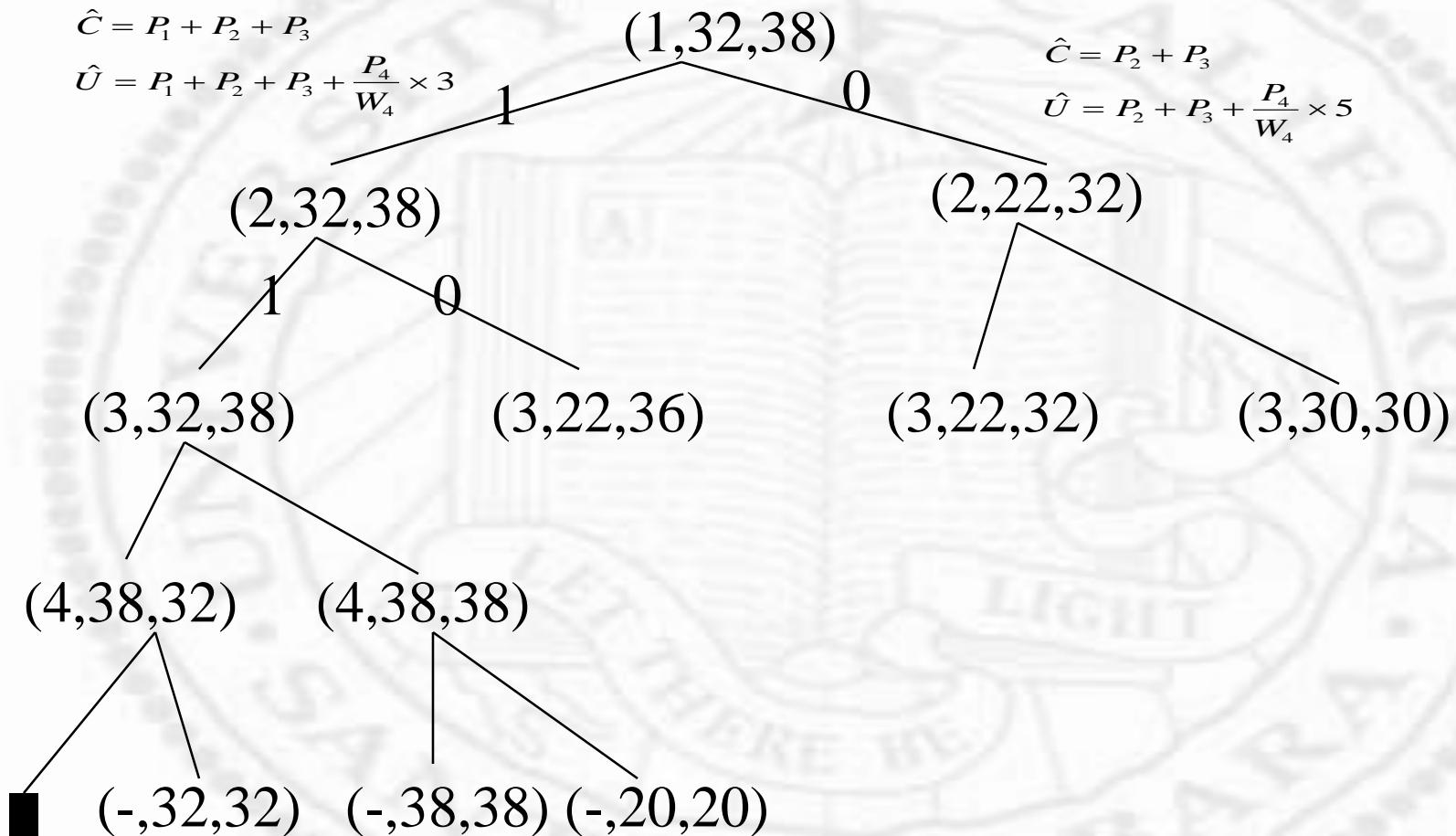
$$k+1, \hat{C}' = \hat{C}, \hat{U}' = \hat{U}$$

$k+1, \hat{C}'$  from integer,  $\hat{U}'$  from greedy



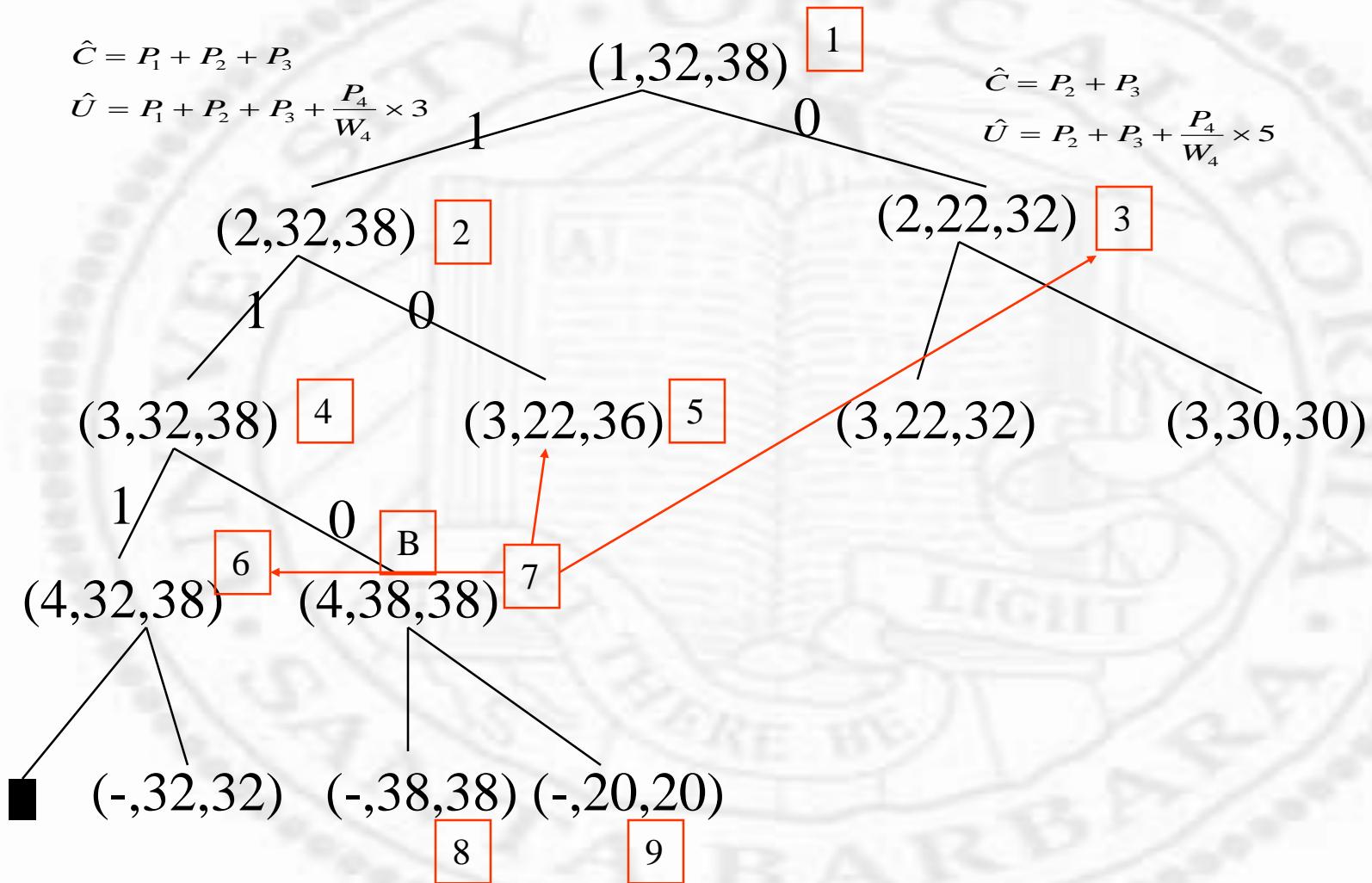
$n=4$   $W=(2, 4, 6, 9)$

$M=15$   $P=(10,10,12,18)$



$n=4$   $W=(2, 4, 6, 9)$

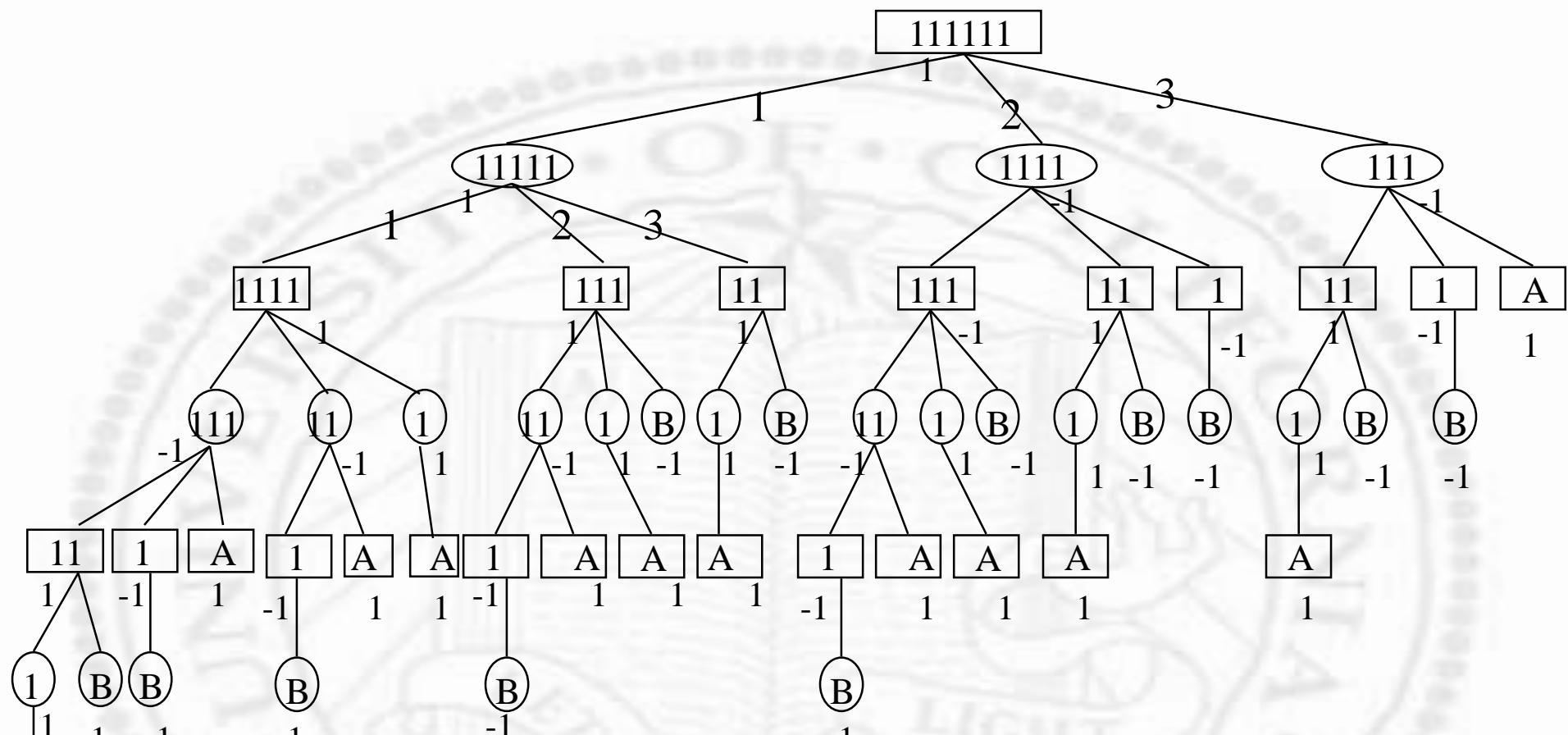
$M=15$   $P=(10,10,12,18)$



# Branch-and-Bound in Game Trees

- Explore many possibilities
- Evaluate their relative merits
- Look ahead finite steps
- *Opposite goals at alternate steps*





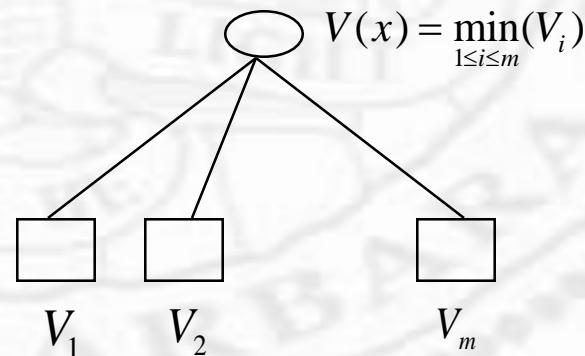
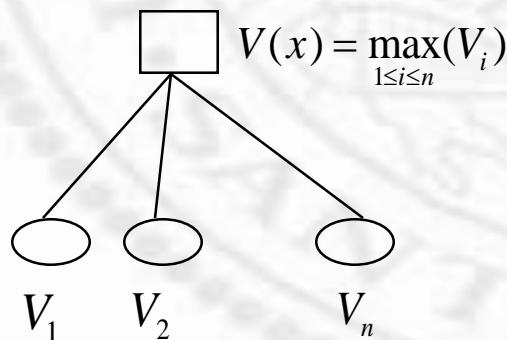
  A's moves  
  B's moves

- Player removes 1 to 3 sticks
- whoever removes the last one loses
- 1 if A wins, -1 if A loses



- Game tree: all possible instances of a finite games
- Game instance: a particular branch from root to a terminal board configuration
- Terminal configuration:

$$V(x) = \begin{cases} 1 & \text{if } x \text{ is a winning configuration for A} \\ 0 & \text{if } x \text{ is a draw configuration for A} \\ -1 & \text{if } x \text{ is a losing configuration for A} \end{cases}$$

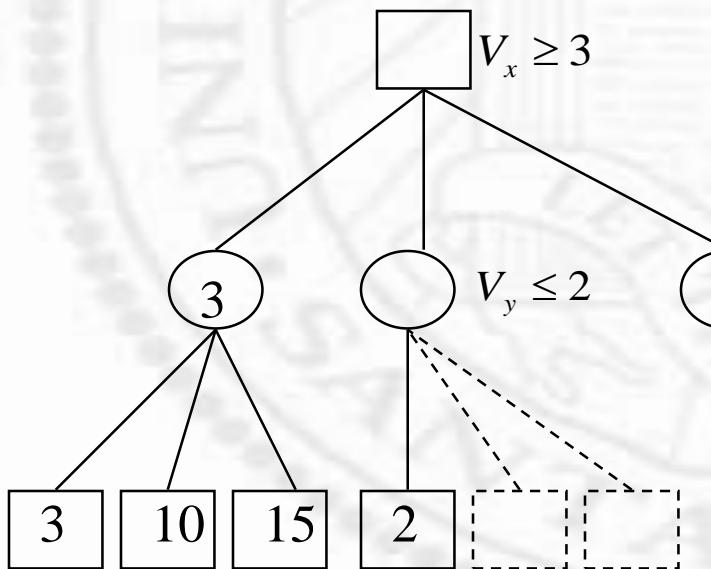


- For chess
  - not possible to generate the whole game tree
  - only a few levels can be generated and evaluated
  - an intelligent evaluation strategy is needed
  - the move strategy is still the same
    - ◆ A to maximize the value
    - ◆ B to minimize the value



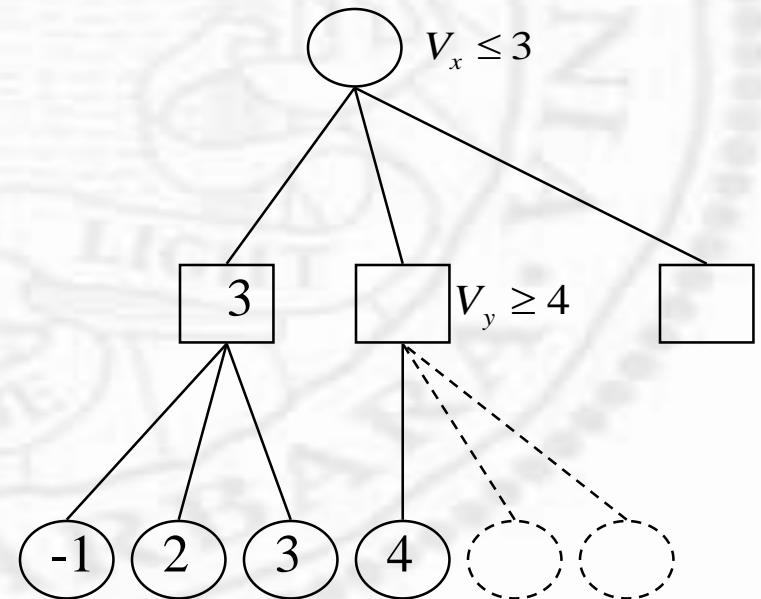
## □ Alpha cutoff

- min V at max position
- if the value of a min position is less than the parent's alpha, the min position need not be explored further

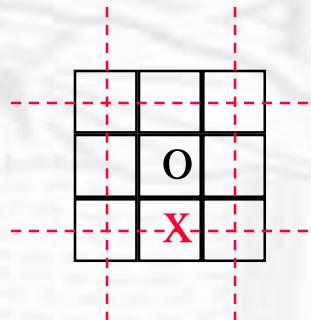
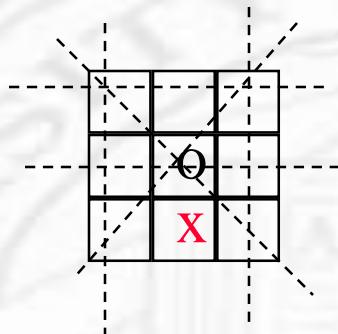


## □ Beta cutoff

- max V at min position
- if the value of a max position is greater than the parent's beta, the max position need not be explored further

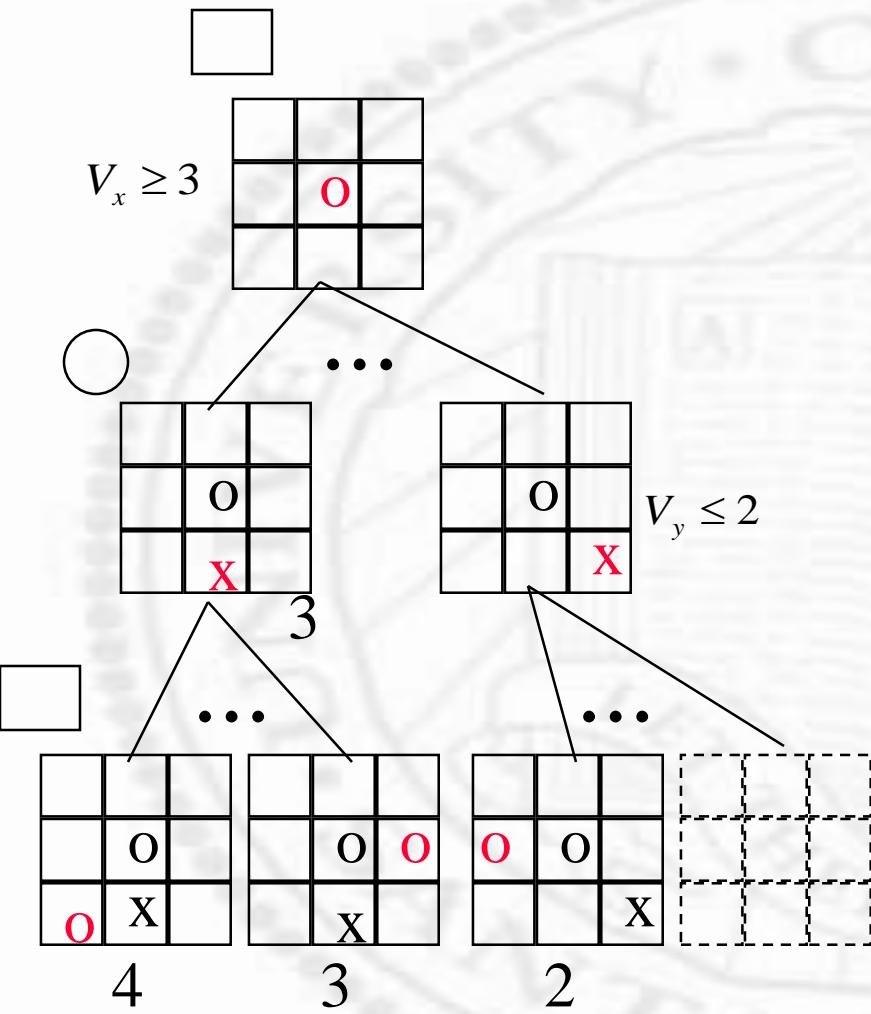


- Example tic-tac-toe
  - evaluation function

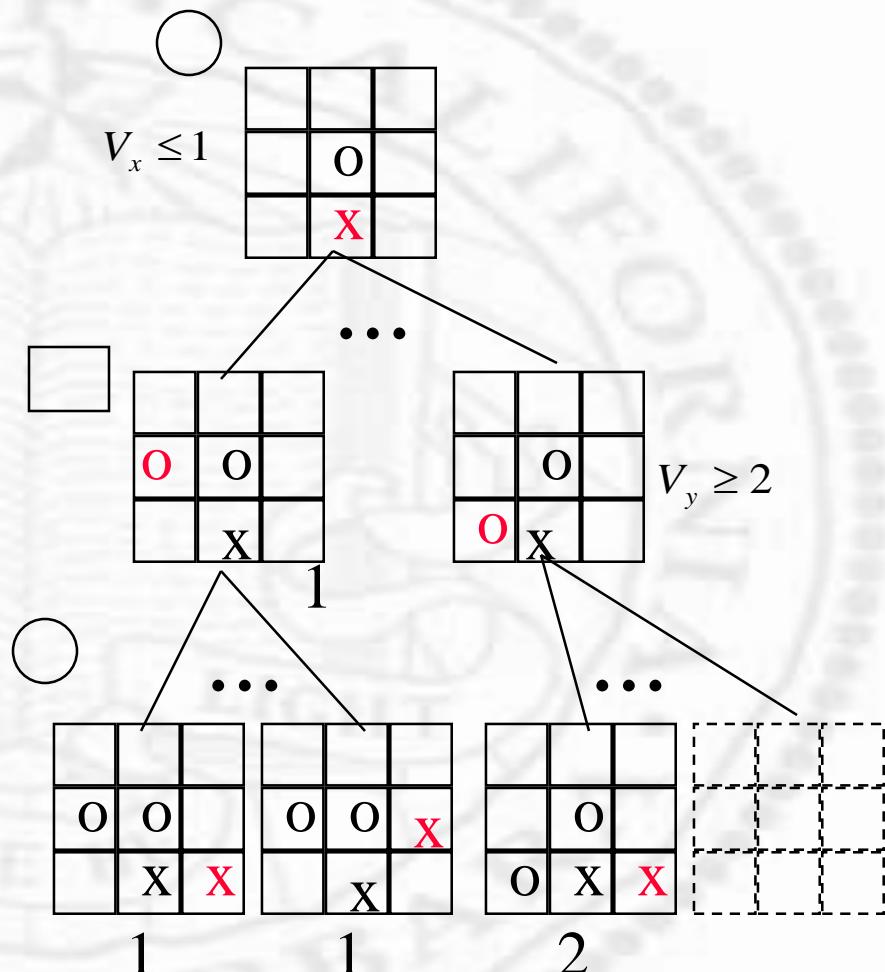


- 6 ways to win for (0) - 4 ways to win for (x)=2 in (0)'s favor

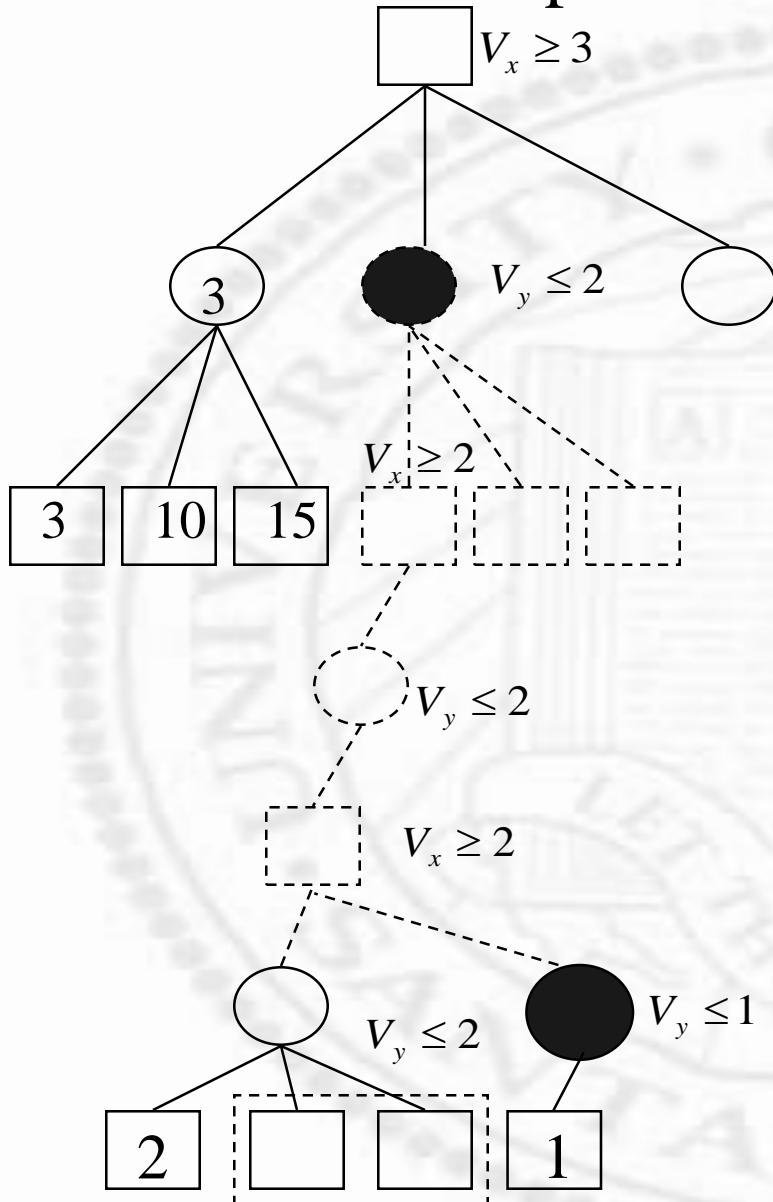
## □ Alpha cutoff



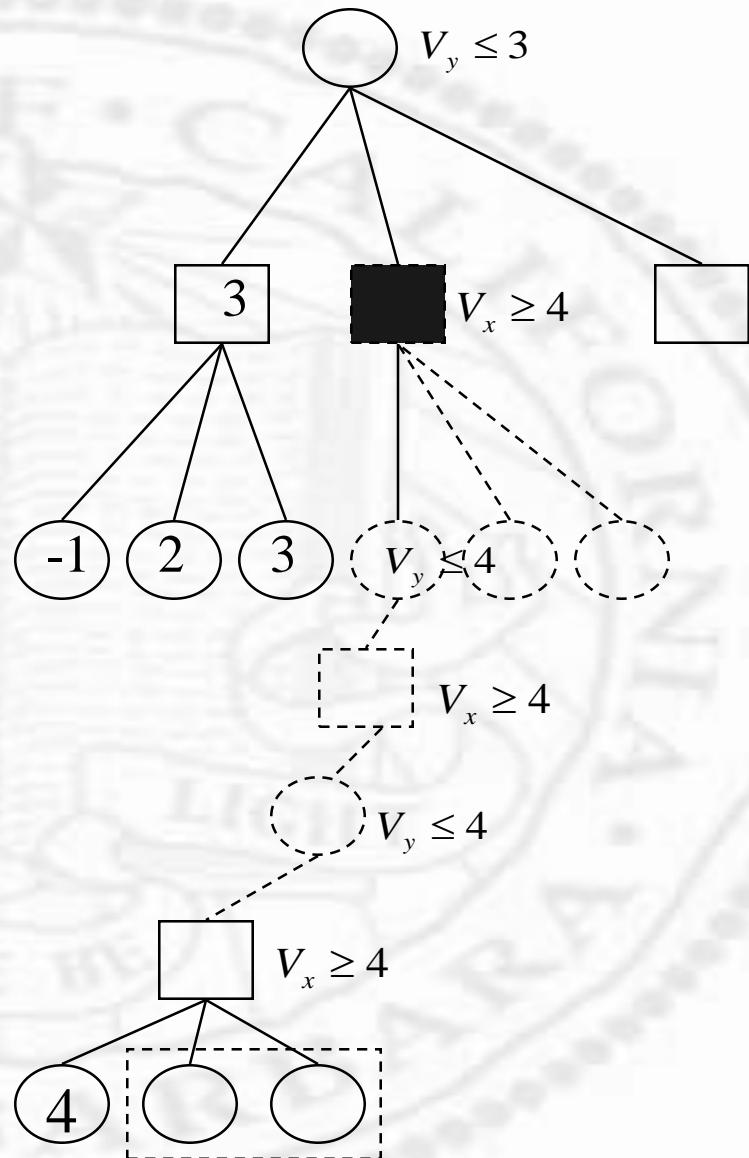
## □ Beta cutoff



## □ Multi-level Alpha cutoff



## □ Multi-level Beta cutoff



Black: bounded