# CSC 421: Algorithm Design & Analysis Spring 2016

#### Backtracking

- greedy vs. backtracking (DFS)
- greedy vs. generate & test
- examples:

N-queens, 2-D gels, Boggle

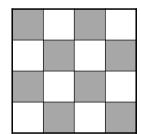
- branch & bound
- problem characteristics
- aside: game trees

1

# Greed is good?

# IMPORTANT: the greedy approach is not applicable to all problems

 but when applicable, it is very effective (no planning or coordination necessary)



GREEDY approach for N-Queens: start with first row, find a valid position in current row, place a queen in that position then move on to the next row

since queen placements are not independent, local choices do not necessarily lead to a global solution

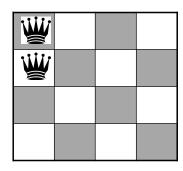
GREEDY does not work - need a more holistic approach

#### Generate & test?

# recall the generate & test solution to N-queens

- systematically generate every possible arrangement
- test each one to see if it is a valid solution

 $\binom{16}{4}$  = 1,820 arrangements



# fortunately, we can do better if we recognize that choices can constrain future choices

- e.g., any board arrangement with a queen at (1,1) and (2,1) is invalid
- no point in looking at the other queens, so can eliminate 16 boards from consideration
- similarly, queen at (1,1) and (2,2) is invalid, so eliminate another 16 boards

3

#### Backtracking

#### backtracking is a smart way of doing generate & test

- view a solution as a sequence of choices/actions (similar to GREEDY)
- when presented with a choice, pick one (*similar to GREEDY*)
- however, reserve the right to change your mind and backtrack to a previous choice (unlike GREEDY)
- you must remember alternatives:
   if a choice does not lead to a solution, back up and try an alternative
- eventually, backtracking will find a solution or exhaust all alternatives

#### backtracking is essentially depth first search

- add ability to prune a path as soon as we know it can't succeed
- when that happens, back up and try another path

# N-Queens psuedocode

```
^{\star} Fills the board with queens starting at specified row
* (Queens have already been placed in rows 0 to row-1)
private boolean placeQueens(int row) {
                                                       if row > board size, then all queens
   if (ROW EXTENDS BEYOND BOARD) {
                                                       have been placed already - return
       return true;
       for (EACH COL IN ROW) {
           if ([ROW][COL] IS NOT IN JEOPARDY FROM EXISTING QUEENS) {
               ADD QUEEN AT [ROW][COL]
                                                       place a queen in available column
               if (this.placeQueens(row+1)) {
                   return true;
                                                       if can recursively place the
                                                       remaining queens, then done
               else {
                   REMOVE QUEEN FROM [ROW][COL]
                                                       if not, remove the queen just placed
                                                       and continue looping to try other
       return false;
                                                       columns
                                                       return false if cannot place
                                                       remaining queens
```

### Chessboard class

#### we could define a class hierarchy for chess pieces

- ChessPiece is an abstract class that specifies the common behaviors of pieces
- Queen, Knight, Pawn, ... are derived from ChessPiece and implement specific behaviors

```
ChessPiece

King Queen Bishop Knight Rook Pawn
```

```
public class ChessBoard {
                                                                   // 2-D array of chess pieces
  private ChessPiece[][] board;
  private int pieceCount;
                                                                  // number of pieces on the board
  public ChessBoard(int size) {...}
                                                                  //\ {\tt constructs}\ {\tt size-by-size}\ {\tt board}
  public ChessPiece get(int row, int col) {...}
public void remove(int row, int col) {...}
                                                                  // returns piece at (row,col)
// removes piece at (row,col)
  public void add(int row, int col, ChessPiece p) {...} // places a piece, e.g., a queen,
                                                                       at (row,col)
  public boolean inJeopardv(int row, int col) {..}
                                                                  // returns true if (row,col) is
                                                                  // under attack by any piece
// returns number of pieces on board
  public int numPieces() {...}
                                                                  // returns the board size
  public int size() {...}
                                                                  // converts board to String
  public String toString() {...}
```

# **Backtracking N-queens**

in an NQueens class, will have a ChessBoard field and a method for placing the queens

 placeQueens calls a helper method with a row # parameter

BASE CASE: if all queens have been placed, then done.

OTHERWISE: try placing queen in the row and recurse to place the rest

note: if recursion fails, must remove the queen in order to backtrack

7

# Why does backtracking work?

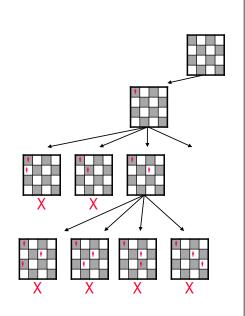
backtracking burns no bridges – all choices are reversible

backtracking provides a systematic way of trying all paths (sequences of choices) until a solution is found

> assuming the search tree is finite, will eventually find a solution or exhaust the entire search space

backtracking is different from generate & test in that choices are made sequentially

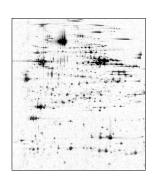
- earlier choices constrain later ones
- can avoid searching entire branches



## Another example: blob count

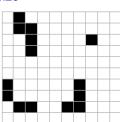
#### application: 2-D gel electrophoresis

- biologists use electrophoresis to produce a gel image of cellular material
- each "blob" (contiguous collection of dark pixels) represents a protein
- identify proteins by matching the blobs up with another known gel image



#### we would like to identify each blob, its location and size

- location is highest & leftmost pixel in the blob
- size is the number of contiguous pixels in the blob
- in this small image: Blob at [0][1]: size 5
  Blob at [2][7]: size 1
  - Blob at [6][0]: size 4 Blob at [6][6]: size 4
- can use backtracking to locate & measure blobs



9

#### Blob count (cont.)

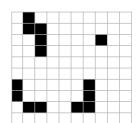
# can use recursive backtracking to get a blob's size when find a spot:

1 (for the spot) +

size of all connected subblobs (adjacent to spot)

#### note: we must not double count any spots

- when a spot has been counted, must "erase" it
- keep it erased until all blobs have been counted



pseudocode:

# Blob count (cont.)

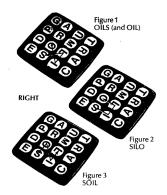
findBlobs traverses the image, checks each grid pixel for a blob

blobSize uses backtracking to expand in all directions once a blob is found

each pixel is "erased" after it is processed in blobSize to avoid double-counting (& infinite recursion)

the image is restored at the end of findBlobs

# Another example: Boggle



#### recall the game

- random letters are placed in a 4x4 grid
- want to find words by connecting adjacent letters (cannot reuse the same letter)
- for each word found, the player earns points
   length of the word
- the player who earns the most points after 3 minutes wins

how do we automate the search for words?

# Boggle (cont.)

can use recursive backtracking to search for a word when the first letter is found:

remove first letter & recursively search for remaining letters again, we must not double count any letters

must "erase" a used letter, but then restore for later searches

G	Α	J	Т
Р	R	М	R
D	0	L	Α
Е	S	I	С

#### pseducode:

13

## BoggleBoard class

# can define a BoggleBoard class that represents a board

- has public method for finding a word
- it calls the private method that implements recursive backtracking
- also needs a constructor for initializing the board with random letters
- also needs a toString method for easily displaying the board

# BoggleGame class

# a separate class can implement the game functionality

- constructor creates the board and fills unguessedWords with all found words
- makeGuess checks to see if the word is valid and has not been guessed, updates the sets accordingly
- also need methods for accessing the guessedWords, unguessedWords, and the board (for display)

**See** BoggleGUI

```
public class BoggleGame {
  private final static String DICT_FILE = "dictionary.txt";
  private BoggleBoard board;
  private Set<String> guessedWords;
  private Set<String> unguessedWords;
  public BoggleGame() {
      this.board = new BoggleBoard();
      this.guessedWords = new TreeSet<String>();
this.unguessedWords = new TreeSet<String>();
          Scanner dictFile = new Scanner(new File(DICT_FILE));
          while (dictFile.hasNext()) {
              String nextWord = dictFile.next();
              if (this.board.findWord(nextWord)) {
                   this.unguessedWords.add(nextWord);
      catch (java.io.FileNotFoundException e) {
          System.out.println("DICTIONARY FILE NOT FOUND");
  public boolean makeGuess(String word) {
      if (this.unguessedWords.contains(word)) {
          this.unguessedWords.remove(word);
          this.guessedWords.add(word);
      return false;
                                                                  15
```

#### **Branch & bound**

the central idea of backtracking is cutting off a branch of the search as soon as we see that it can't lead to a solution

• then, backtrack and try a different branch

#### e.g., for the shortest path problem

- we cut off a branch if the vertex was a dead end
- we also cut it off if its length exceeded that of an already found path

#### what if we also had the ability to look ahead?

- i.e., if we could tell ahead of time (using some deduction) that a branch was not going to work, then we could preemptively cut
- this variant of backtracking is known as branch & bound

# B & B example

suppose you have four jobs and 4 contractors (with bids), and want to assign the jobs to the contractors to minimize cost

	job 1	job 2	job 3	job 4
contractor a	\$9K	\$2K	\$7K	\$8K
contractor b	\$6K	\$4K	\$3K	\$7K
contractor c	\$5K	\$8K	\$1K	\$8K
contractor d	\$7K	\$6K	\$9K	\$4K

- e.g.,  $a \rightarrow 1$ ,  $b \rightarrow 2$ ,  $c \rightarrow 3$ ,  $d \rightarrow 4$
- 9 + 4 + 1 + 4 = \$18K total
- e.g.,  $a \rightarrow 2$ ,  $b \rightarrow 3$ ,  $c \rightarrow 1$ ,  $d \rightarrow 4$
- 2 + 3 + 5 + 4 = \$14K total

generate & test?

17

# B & B example (cont.)

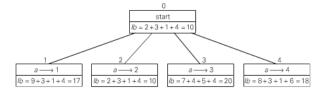
	job 1	job 2	job 3	job 4
contractor a	\$9K	\$2K	\$7K	\$8K
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contractor c	\$5K	\$8K	\$1K	\$8K
contractor d	\$7K	\$6K	\$9K	\$4K

#### note that there is a (possibly unobtainable) lower bound on the bid total

- you can't possibly do better than assigning every contractor their lowest bid
- here, 2 + 3 + 1 + 4 = \$10K is a lower bound
- (it is not even achievable, since b & c are assigned the same job)

#### the lower bound gives us a basis for choosing one branch over another

• i.e., use a greedy approach to select the branch with smallest lower bound lb = cost of bids assigned so far + minimal bids possible for remaining contractors

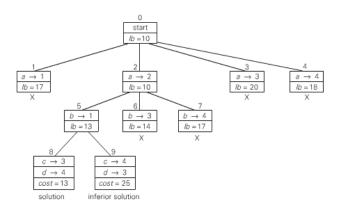


# B & B example (cont.)

	job 1	job 2	job 3	job 4
contractor a	\$9K	\$2K	\$7K	\$8K
contractor b	\$6K	\$4K	\$3K	\$7K
contractor c	\$5K	\$8K	\$1K	\$8K
contractor d	\$7K	\$6K	\$9K	\$4K

#### at each step, choose the vertex/state with smallest lower bound, and extend

can cut off a branch if its lb exceeds the cost of a found solution



19

# Interesting aside: B & B search in game playing

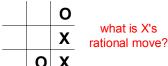
consider games involving:

- 2 players
- perfect information
- zero-sum (player's gain is opponent's loss)

examples: tic-tac-toe, checkers, chess, othello, ... non-examples: poker, backgammon, prisoner's dilemma, ...

von Neumann (the father of game theory) showed that for such games, there is always a "rational" strategy

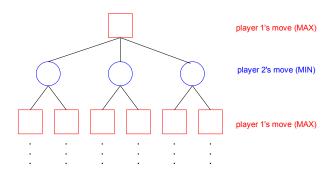
 that is, can always determine a best move, assuming the opponent is equally rational



#### Game trees

#### idea: model the game as a search tree

- associate a value with each game state (possible since zero-sum) player 1 wants to maximize the state value (call him/her MAX) player 2 wants to minimize the state value (call him/her MIN)
- players alternate turns, so differentiate MAX and MIN levels in the tree



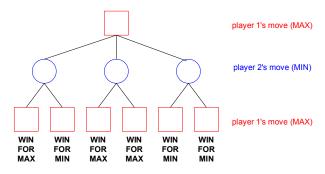
the leaves of the tree will be end-of-game states

2

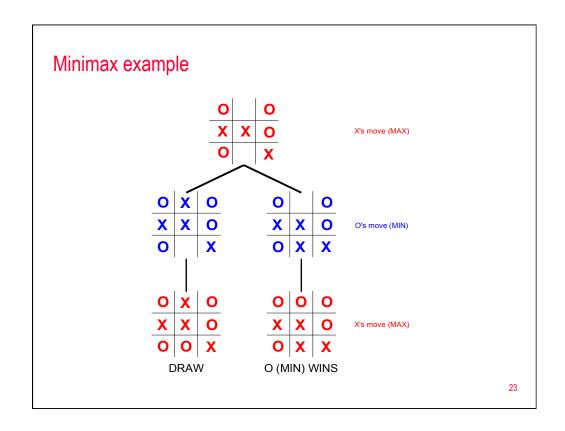
#### Minimax search

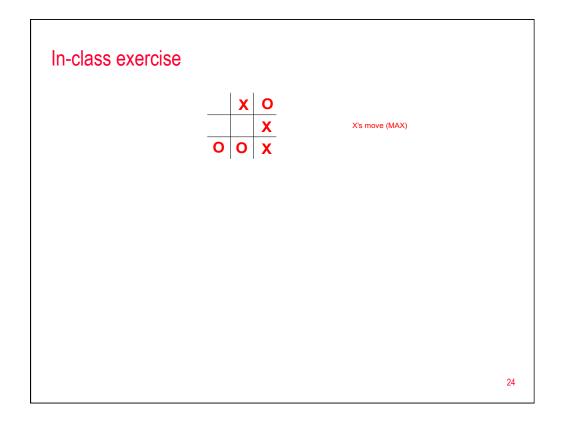
#### minimax search:

- at a MAX level, take the maximum of all possible moves
- at a MIN level, take the minimum of all possible moves



can visualize the search bottom-up (start at leaves, work up to root) likewise, can search top-down using recursion





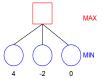
# Minimax in practice

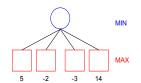
while Minimax Principle holds for all 2-party, perfect info, zero-sum games, an exhaustive search to find best move may be infeasible

EXAMPLE: in an average chess game, ~100 moves with ~35 options/move → ~35<sup>100</sup> states in the search tree!

#### practical alternative: limit the search depth and use heuristics

- expand the search tree a limited number of levels (limited look-ahead)
- evaluate the "pseudo-leaves" using a heuristic high value → good for MAX low value → good for MIN back up the heuristic estimates to determine the best-looking move at MAX level, take maximum at MIN level, take minimum



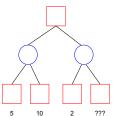


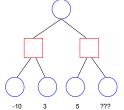
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# Tic-tac-toe example heuristic(State) = 1000 -1000 (#rows/cols/diags open for MAX – #rows/cols/diags open for MIN) suppose look-ahead of 2 moves 1000 -1000 (#rows/cols/diags open for MIN) suppose look-ahead of 2 moves

# $\alpha\text{-}\beta$ bounds

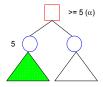
sometimes, it isn't necessary to search the entire tree

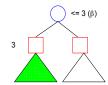




#### $\alpha$ - $\beta$ technique: associate bonds with state in the search

- associate lower bound  $\alpha$  with MAX: can increase
- associate upper bound  $\beta$  with MIN: can decrease

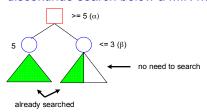


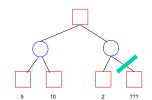


27

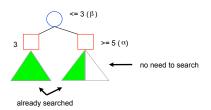
# $\alpha$ - $\beta$ pruning

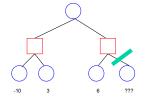
discontinue search below a MIN node if  $\beta$  value <=  $\alpha$  value of ancestor

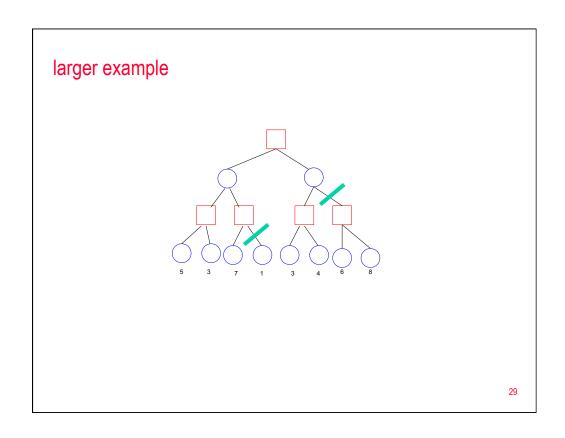


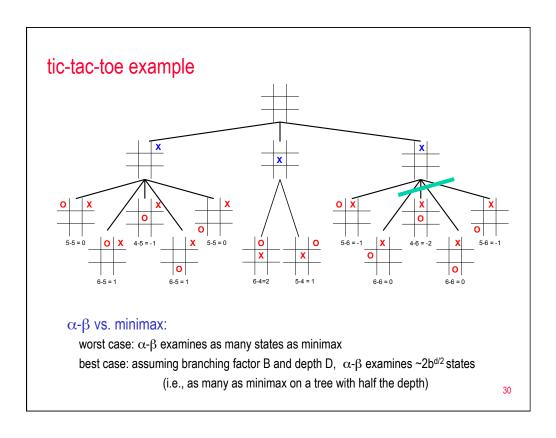


#### discontinue search below a MAX node if $\alpha$ value >= $\beta$ value of ancestor









# Iterative deepening

#### a common approach in game search is to set a lookahead range

- e.g., in chess, lookahead 4 moves (by each player) and rate boards at that stage
- this catches wins/losses within that range
- presumably, you can better judge the state of the game in the future

#### if decisions are timed.

- you can pick a conservative lookahead range to ensure a choice is made
- if time remains, extend the lookahead range and try again
- each iteration looks deeper and so makes a more informed choice

#### clearly, there is redundancy with iterative deepening

- lookead search of (n+1) levels must reproduce the search for n levels
- HOW COSTLY IS THIS?

3

## Algorithmic approaches summary (so far)

brute force: sometimes the straightforward approach suffices

transform & conquer: sometimes the solution to a simpler variant suffices

divide/decrease & conquer: tackles a complex problem by breaking it into smaller piece(s), solving each piece (often w/ recursion), and combining into an overall solution

applicable for any application that can be divided into smaller or independent parts

greedy: makes a sequence of choices/actions, choose whichever looks best at the moment

applicable when a solution is a sequence of moves & perfect knowledge is available

backtracking: makes a sequence of choices/actions (similar to greedy), but stores alternatives so that they can be attempted if the current choices lead to failure

- more costly in terms of time and memory than greedy, but general-purpose
- branch & bound variant cuts off search at some level and backtracks