# Lecture 22: Dynamic Programming IV

### Lecture Overview

- 2 kinds of guessing
- Piano/Guitar Fingering
- Tetris Training
- Super Mario Bros.

## **Review:**

\* 5 easy steps to dynamic programming

(a) define subproblems	${\rm count}\ \#\ {\rm subproblems}$
(b) guess (part of solution)	count $\#$ choices
(c) relate subproblem solutions	compute time/subproblem
<ul> <li>(d) recurse + memoize</li> <li>problems</li> <li>OR build DP table bottom-up</li> <li>check subproblems acyclic/topological order</li> </ul>	time = time/subproblem $\cdot \#$ sub-
(e) solve original problem: = a subproblem OR by combining subproblem solutions	$\implies$ extra time

\* 2 kinds of guessing:

- (A) In (3), guess which other subproblems to use (used by every DP except Fibonacci)
- (B) In (1), create more subproblems to guess/remember more structure of solution used by knapsack DP
  - effectively report many solutions to subproblem.
  - lets parent subproblem know features of solution.

## Piano/Guitar Fingering:

## Piano

[Parncutt, Sloboda, Clarke, Raekallio, Desain, 1997][Hart, Bosch, Tsai 2000][Al Kasimi, Nichols, Raphael 2007] etc.

• given musical piece to play, say sequence of n (single) notes with right hand

- fingers  $1, 2, \ldots, F = 5$  for humans
- metric d(f, p, g, q) of difficulty going from note p with finger f to note q with finger g

e.g.,  $1 < f < g \& p > q \implies$  uncomfortable stretch rule:  $p \ll q \implies$  uncomfortable legato (smooth)  $\implies \infty$  if f = gweak-finger rule: prefer to avoid  $g \in \{4, 5\}$  $3 \rightarrow 4 \& 4 \rightarrow 3$  annoying  $\sim$  etc.

### First Attempt:

- 1. subproblem = min. difficulty for suffix notes [i:]
- 2. guessing = finger f for first note [i]
- 3. recurrence:  $DP[i] = \min(DP[i+1] + d(\text{note}[i], f, \text{note}[i+1], ?) \text{ for } f \cdots)$   $\rightarrow \text{ not enough information!}$

### Correct DP:

- 1. <u>subproblem</u> = min difficulty for suffix notes[i :] given finger f on first note[i]  $\implies n \cdot F$  subproblems
- 2. <u>guessing</u> = finger g for next note[i + 1]  $\implies F$  choices
- 3. <u>recurrence</u>:  $DP[i, f] = \min(DP[i+1, g] + d(\text{note}[i], f, \text{note}[i+1], g) \text{ for } g \text{ in range}(F))$  DP[n, f] = 0 $\implies \Theta(F) \text{ time/subproblem}$
- 4. topo. order: for *i* in reversed(range(*n*)): for *f* in 1, 2, ..., *F*: total time  $O(nF^2)$
- 5. <u>orig. prob.</u> = min(DP[0, f] for f in 1, ..., F) (guessing very first finger)

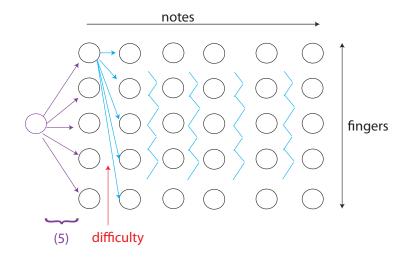


Figure 1: DAG.

## Guitar

Up to S ways to play same note! (where S is # strings)

- redefine "finger" = finger playing note + string playing note
- $\bullet \implies F \to F \cdot S$

### Generalization:

Multiple notes at once e.g. chords

- input: notes[i] = list of  $\leq F$  notes (can't play > 1 note with a finger)
- <u>state</u> we need to know about "past" now assignment of F fingers to  $\leq F+1$  notes/null  $\implies (F+1)^F$  such mappings
- (1)  $n \cdot (F+1)^F$  subproblems where  $(F+1)^F$  is how notes [i] is played
- (2)  $(F+1)^F$  choices (how notes[i+1] played)
- (3)  $n \cdot (F+1)^{2F}$  total time
  - works for 2 hands F = 10
  - just need to define appropriate d



Figure 2: Tetris.

## Tetris Training:

- given sequence of n Tetris pieces & an empty board of small width w
- must choose orientation & x coordinate for each
- then must drop piece till it hits something
- full rows do not clear without the above two artificialities WE DON'T KNOW! (but: if nonempty board & w large then NP-complete)
- goal: survive i.e., stay within height h

### First Attempt:

- 1. subproblem = survive in suffix i? WRONG
- 2. guessing = how to drop piece  $i \implies \#$  choices = O(w)
- 3. recurrence: DP[i] = DP[i+1] ?! not enough information! What do we need to know about prefix : *i*?

#### **Correct:**

- 1. <u>subproblem</u> = survive? in suffix *i*: given initial column occupancies  $h_0, h_1, \dots, h_{w-1}$ , call it  $h \implies \#$  subproblems =  $O(n \cdot h^w)$
- 3. <u>recurrence</u>:  $DP[i, h] = \max(DP[i, m] \text{ for valid moves } m \text{ of piece } i \text{ in } h)$  $\implies$  time per subproblem = O(w)
- 4. <u>topo. order</u>: for *i* in reversed(range(n)): for  $h \cdots$ total time =  $O(nwh^w)$  (DAG as above)
- 5. <u>solution</u> = DP[0, 0]
   (& use parent pointers to recover moves)

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## Super Mario Bros

Platform Video Game

- given entire level (objects, enemies, ...)  $(\leftarrow n)$
- small  $w \times h$  screen
- configuration
  - screen shift  $(\leftarrow n)$
  - player position & velocity  $(O(1)) (\leftarrow w)$
  - object states, monster positions, etc. ( $\leftarrow c^{w.\cdot h}$ )
  - anything outside screen gets reset ( $\leftarrow c^{w.\cdot h}$ )
  - score ( $\leftarrow S$ )
  - time ( $\leftarrow T$ )
- transition function δ: (config, action) → config' nothing, ↑, ↓, ←, →, B, A press/release
- (1) <u>subproblem</u>: best score (or time) from config. C  $\implies n \cdot c^{w \cdot h} \cdot S \cdot T$  subproblems
- (2) <u>guess</u>: next action to take from C  $\implies O(1)$  choices
- (3) <u>recurrence</u>:

$$DP(C) = \begin{cases} C.\text{score} & \text{if on flag} \\ \infty & \text{if } C.\text{dead or } C.\text{time} = \\ \max(DP(\delta(C, A))) & \text{for } A \text{ in actions} \end{cases}$$

 $\implies O(1)$  time/subproblem

(4) topo. order: increasing time

- (5) orig. prob.: DP(start config.)
  - pseudopolynomial in S & T
  - polynomial in n
  - exponential in  $w \cdot h$

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6.006 Introduction to Algorithms Fall 2011

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