Monday, March 28-Day 25 \bigcirc Topic 10-Introduction to Metaheuristics So far, we've mostly focused on ways to find optimal solutions. These techniques are <u>slow</u>, and <u>difficult</u>, aren't always applicable. Ex: TSP, brute force takes O(n!) time dynamic programming - $O(n^2 \cdot 2^n)$ time Metaheuristics: - General problem solving paradigms that can easily adapted to many problems - Look for good solutions, and rarely actually find on optimal one. - Roethy find - Yretty fast Similar set up: made of candidates/solutions has some <u>Score</u>/fitness /guality * Search Space * Every Solution

* Goal: Find a candidate with a good score Lassume always maximizing unless specified otherwise) score Many of our problems will be discrete (finite search space), but some will be continuous. Ex: Find the maximum value of $f(x) = \frac{1}{\cos(\log(x+1)^2)} \cdot \sin(\min((x+1)^{100}, \frac{1}{x}))$ on the interval $0.02 \leq x \leq 0.04$. Also work on functions that are defined implicity. Search space: set of real #5 from U.OZ to 0.04. Most of our search spaces will not be 1 dimensional.

Sin 2 (x-y) Sm 2 (x+y) Vy2+y2 0.5 Goal: find the top of the tallest hill $\int_{-2\pi} \frac{3\pi}{2} -\pi -\frac{\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\pi}{2} \frac{3\pi}{2} \frac{2\pi}{2} \int_{-2\pi} \frac{3\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\pi}{2} \frac{3\pi}{2} \frac{2\pi}{2} \int_{-2\pi} \frac{3\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\pi}{2} \frac{3\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\pi}{2} \frac{3\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\pi}{2} \frac{3\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\sigma}{2} \frac{\pi}{2} \frac{\sigma}{2} \frac{\sigma}{2$ 5 without getting stuck on the wrong hilltop Rondoms ~ 21 Random greedy: ≈ 7.12 HC $\propto 6.71$ × 6.299

Gradient Ascent (or Descent) * Optimization method you learn in some classes * If your function f(xy) is differentiable, you can compute the gradient at any point. Gradient is a vector that points in the direction of steepost ascent.

5A

(i) Start at a point (c) compute the gradient at that point (3) move a little bit in that divection (4) repeat



0.6 0.5 0,4 0.3 0.2 0.1 0.8 1.2 , 1.4 1.6 1.0 1.8 2.0 0.5 1.5 Will lead to a local max, but not a global max. Might overshoot. How could you simulate Gradient Ascent with a discrete search space. continuous [pretend you're standing in the mountains]

* start at a random point ★ look around yourself in a small radius (plug in some points near the current point, and see which is highest) * go there and repeat Discuete, TSP as an example * standing at some possible solution (some tour) * look at other tours "near" the current tour * move to the best one and repeat - need a definition of "near by"/"small radius" 5 cities, 1, 2, 3, 4, 5 At the solution: 3 - 5 -> 2 -> 1 -> 4 -> 3 What tours are "close" to this one? One possibility: nearby = switch two consecutive cities Others: