

Wed, Apr. 3, 2024
Scientific Computing

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Announcements

- * Homework 4 due Fri (or Mon if you have asked for an extension)
- * No in-person lecture on Mon, I will post a video
- No O.H. on Monday.

Topic 11 - Hill-Climbing

RECAP

Problem Setup:

- * Search space S of candidates
- * Scoring function: $\text{Score}(x)$ for $x \in S$
(also called fitness or quality)

- * A way to generate either:
 - all the candidates "near" a particular candidate

the set of all candidates "near x " is called the "neighborhood of x "
Notation: $\text{nbhd}(x)$

OR

- a random candidate "near" another one, a "tweak"
 $\text{tweak}(x)$ = an element of the search space that is "near" x .

- "nearby" is up for you to decide
- different definitions can give better or worse solutions

Two running examples in this lecture:

(1) TSP:

* discrete

* search space = all tours of the cities

* score = sum of the distances traveled, we are minimizing

* Let $x = C_1 \rightarrow C_2 \rightarrow \dots \rightarrow C_n \rightarrow C_1$

Define $\text{nbhd}(x)$ to be all tours you can get by swapping two cities.

How big is $\text{nbhd}(x)$?

$$\left. \begin{array}{l} A \rightarrow B \rightarrow C \rightarrow D \rightarrow E \rightarrow A \\ A \rightarrow D \rightarrow C \rightarrow B \rightarrow E \rightarrow A \end{array} \right\} \binom{n-1}{2} = \frac{(n-1) \cdot (n-2)}{2}$$

$$\approx n^2/2$$

* tweak(x): randomly pick two cities and swap the

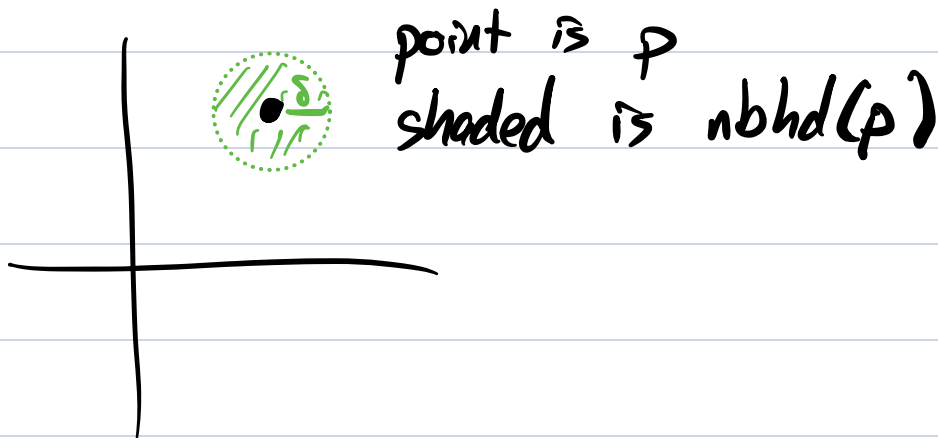


(2) optimizing a continuous function
in two variables $f(x,y)$

* continuous search space containing
all (x,y) points, maybe in some
bounds

* Score = the value of the
function at that point

* $\text{nbhd}(p)$ = all points within some
fixed distance δ of p .
Euclidean



* $\text{tweak}(p)$ = one random point in
 $\text{nbhd}(p)$

Metaheuristic #1: Random Search

best = random element of S

while True:

x = random element of S

 if score(x) > score(best):

 best = x

(quit whenever you want)

Stopping Conditions (many possible):

- * best score hasn't improved for N iterations
- * preset # of iterations to do
- * you get bored
- * if you have some sense ahead of time of what the optimal score should be, stop when you get within ϵ of that score

This not a good MH in most cases because it does not retain information to guide future choices.

[2 demos] { 01: TSP random
 02: Contour-1 random

MH #2 - Steepest Ascent Hill-Climbing

inspired by gradient ascent

(discrete only,
otherwise
 $|N| = \infty$)

$x =$ random element of S

while True:

$N = \text{nbhd}(x)$

$s =$ element of N with the best score

if $\text{score}(s) > \text{score}(x)$:

$x = s$

else:

quit

what if a tie?
up to you

May need stopping conditions like before

This mimics gradient ascent but for discrete spaces. It climbs right up to the top of a hill, then stops.

Pros

* Find a local optimum

Cons

* Unlikely to

* Fast-ish

find a global
optimum except in
very nice spaces
* Slow when nbhds
are big.

↪ Generating and Scoring
big neighborhoods

TSP with 300 cities

Scoring one element of the search
space is not that bad

300 distance calculations

each is two subtractions

two squarings

one addition

one $\sqrt{\quad}$

Size of nbhd: $\binom{299}{2} = 44,551$

Scoring 44,551 of them is slow.

Demos 3, 4