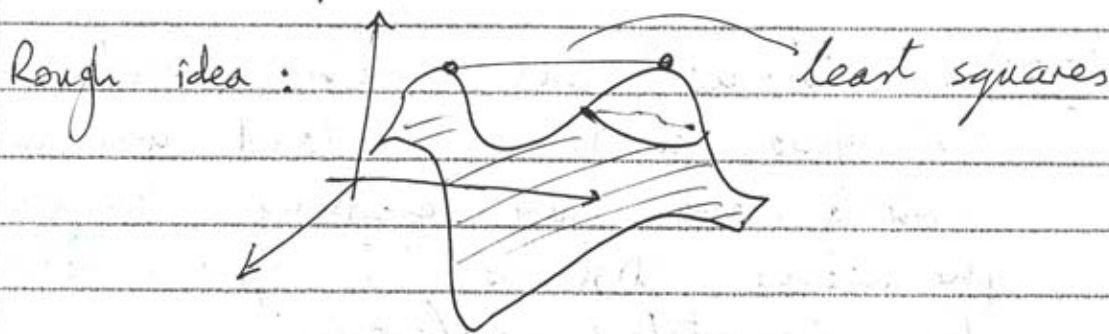


# Lecture Notes

BRIDSON - AIM, JUNE 1, 2003, PALO ALTO

## - DISTANCES, SPACES + CURVATURE

- 1) - Why look at different spaces/distances rather than just doing distance estimates in Euclidean space  $\mathbb{R}^n$ ?



Data involves some measurement, giving points in  $\mathbb{R}^n$ . You want to talk about "CLOSE" data points. Typically, you might do a least squares approximation, which amounts to taking STRAIGHT-LINE DISTANCE in  $\mathbb{R}^n$ .

BUT a more natural measure of SEPARATION (DISTANCE) between data points is the LENGTH of path IN THE ~~DATA~~ SPACE OF POSSIBLE DATA (drawn as a surface).

2.

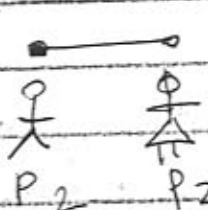
Thus we eschew EXTRINSIC distance (the straight line) for the INTRINSIC distance in the space of outcomes.

- So, we want to work in SPACES with non-Euclidean notions of DISTANCE (separation).

ASIDE: "distance" and intuition/vocabulary of space should not mislead you into thinking we're only describing spatial phenomena. "Distance" is just a metaphor for (quantified) difference.

EXAMPLE: Hand-shaking graph.

VERTICES = PEOPLE ON EARTH

EDGES =  |  $p_1$  has shaken hands with  $p_2$

- Distance - # edges to be traversed in going from one vertex to another in shortest path.

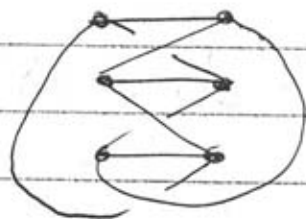
## Points to Observe :

1) We had a notion of length and turned it into a notion of distance (we'll go the other way later).

2) One can embed this graph in  $\mathbb{R}^3$  in a length-preserving way, but the straight-line distances of the embedding will have little or nothing to do with the natural notion of distance defined above.  
— And why bother! Unnatural.

— 3) Although one can embed any graph in  $\mathbb{R}^3$ , one cannot embed in  $\mathbb{R}^2$ , necessarily

e.g. utilities graph  $K_{3,3}$

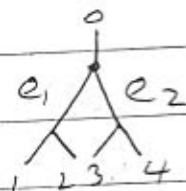


MORAL : As we look at higher dimensional spaces, we can't hope for low-dimensional ( $\leq 4$ ) embeddings

4.

- Another example (seen in previous talk)

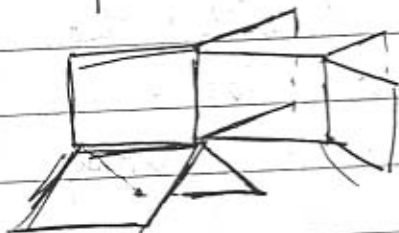
TREE SPACE  
 $n=4$



two edge  
lengths  $|e_1| + |e_2| = 1$

- as internal edges degenerate we get change of topology; as edges vary we move around in a square

- we get a space built out of squares:



can't expect a Euclidean embedding, misleading to try a least-squares distance, but there IS a natural distance:

$d(x, y)$  = length of shortest path from  $x$  to  $y$

measured LOCALLY in squares

GLOBAL measure of separation.

KEY IDEA : Geometry (of spaces with notion of distance) allows one to use LOCAL information to make GLOBAL conclusions. (a process hopeless with discrete spaces — thus we want to arrange our data into spaces)

SLIDE:1 — METRIC SPACES

- Deliberately pared down to essential features
- Restricting to GEODESIC metrics makes geometry possible in a real sense (in particular opens the possibility of a local-to-global analysis of the space).

IMPORTANT REMARKS: ① Can go from notion of distance TO length. (and can go back in geodesic spaces)

② SEPARATION  $\rightsquigarrow$  distance

In many circumstances one may have a notion of separation of data,

6.

$$S: X \times X \rightarrow \mathbb{R} \text{ (or } \mathbb{N})$$

that does not satisfy the triangle inequality. In such circumstances, one can manufacture a metric by

$d(x, y) =$  length of shortest sequence of hops  $x \rightsquigarrow x_1 \rightsquigarrow \dots \rightsquigarrow x_n = y$

such that  $S(x_i, x_{i+1}) \leq \text{constant}$ .

— or some similar process.

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SLIDES 2+3 :

Hyperbolic graphs, with explanation. Explanation of GROWTH part — the difficulty of embedding things in Euclidean space is not just a question of dimension

SLIDE 4 - Geodesics  
- Core issues

Re. comparison: coordinates  $\rightarrow$  Riemannian geometry

comparison of finite ~~sets~~ configurations....

SLIDE 5: CURVATURE

SLIDE 6: CAT(0) spaces

Emphasis - the local-to-global POWER

Where is this power coming from - CONVEXITY

SLIDE 7: CONVEXITY

SLIDE 8: CENTRES

- Explanation of Euclidean origins. Discussion.
- Emphasis on the applications to tree space.

- There follows a detailed description of (a variation on) the link of the star vertex in tree space for  $n=5$