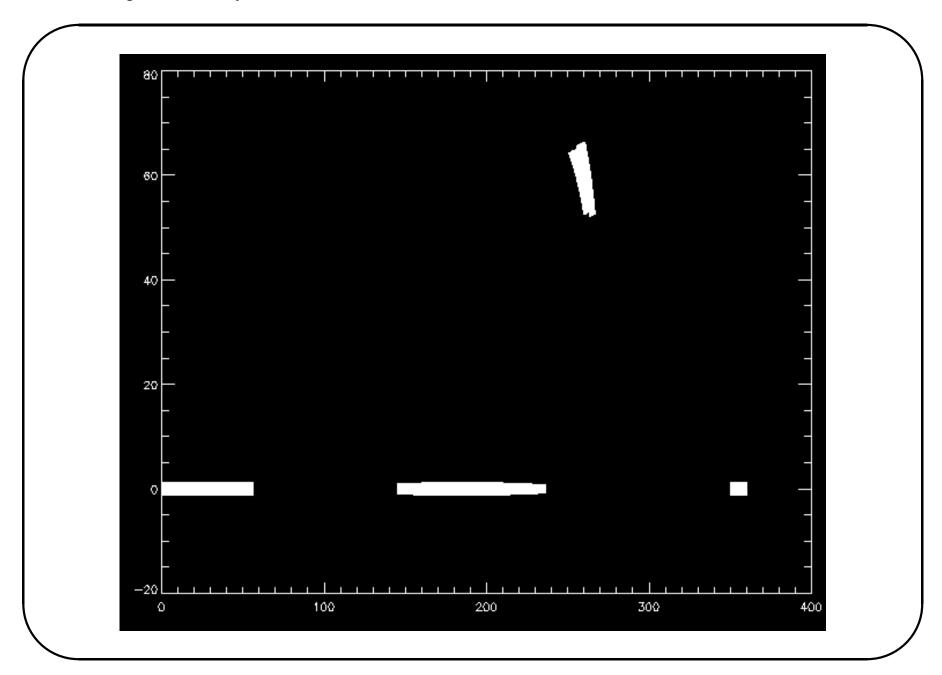
## **Hierarchic Clustering of 3D Galaxy Distributions**

#### Topics:

- Data
- Hierarchic clustering
- Ultrametric topology
- P-adic algebra
- Practical interest
- Testing for ultrametricity
- Lerman's H-classifiability
- Conclusion and critique

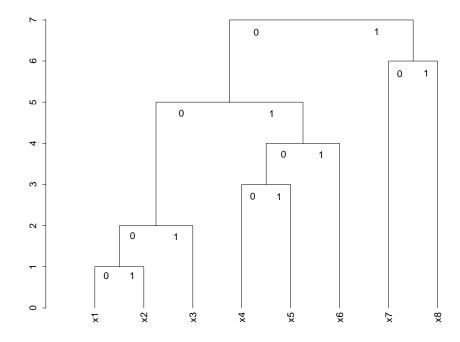
Data

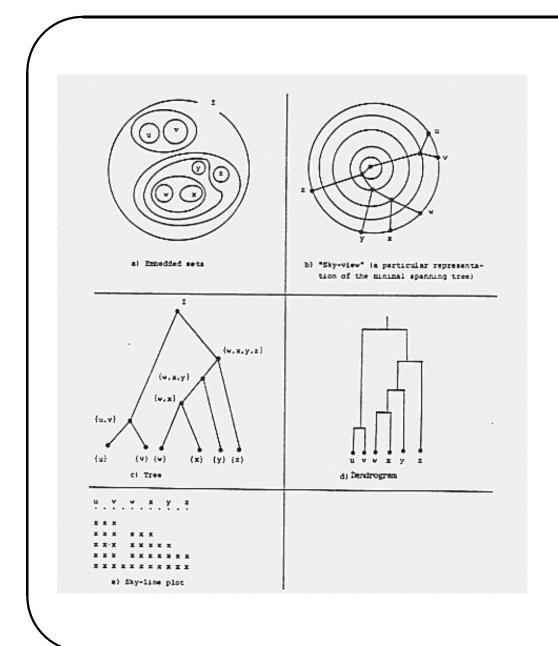
- Sloan Digital Sky Survey data
- RA, Dec, redshift value, reliability indicator
- 345109 galaxies in right ascension and declination, photometric redshift
- In this work we used the low RA, galaxy plane area.



# **Hierarchic Clustering**

Labeled, ranked dendrogram on 8 terminal nodes. Branches labeled 0 and 1.





## **Hierarchic Clustering: Metric** ⇒ **Ultrametric**

- Hierarchical agglomeration on n observation vectors,  $i \in I$
- Series of  $1, 2, \ldots, n-1$  pairwise agglomerations of observations or clusters
- Hierarchy  $H = \{q | q \in 2^I\}$  such that (i)  $I \in H$ , (ii)  $i \in H \ \forall i$ , and (iii) for each  $q \in H, q' \in H: q \cap q' \neq \emptyset \Longrightarrow q \subset q' \text{ or } q' \subset q$ .
- Indexed hierarchy is the pair  $(H, \nu)$  where the positive function defined on H, i.e.,  $\nu: H \to \mathbb{R}^+$ , satisfies:  $\nu(i) = 0$  if  $i \in H$  is a singleton; and (ii)  $q \subset q' \Longrightarrow \nu(q) < \nu(q')$ . Function  $\nu$  is the agglomeration level.
- Take  $q \subset q'$ , let  $q \subset q''$  and  $q' \subset q''$ , and let q'' be the lowest level cluster for which this is true. Then if we define  $D(q, q') = \nu(q'')$ , D is an ultrametric.

## **Ultrametric Spaces and Properties**

- Let (E, d) be a metric space, i.e. a set E and a positive function  $E \times E \longrightarrow \mathbb{R}_+$  satisfying
  - 1. d(x,y) = d(y,x)
  - 2. d(x, y) = 0 iff x = y
  - 3.  $d(x,z) \le d(x,y) + d(y,z)$

A space is ultrametric if in addition we have  $d(x, z) \leq \max(d(x, y), d(y, z))$ 

- A metric space (E, d) is ultrametric iff all its triangles are isosceles, with the length of the base being less than or equal to that of the sides.
- Each point of a circle in E is its center. Each ball in an ultrametric space is both open and closed.
- Two non-disjoint balls are concentric.

## **P-adic Coding**

- For the dendrogram shown in we develop the following p-adic encoding for p=2 of terminal nodes, traversing a path from the root.
- $x_1 = 0 \cdot 2^7 + 0 \cdot 2^5 + 0 \cdot 2^2 + 0 \cdot 2^1$ ;
- $x_2 = 0 \cdot 2^7 + 0 \cdot 2^5 + 0 \cdot 2^2 + 1 \cdot 2^1$ ;
- $x_4 = 0 \cdot 2^7 + 1 \cdot 2^5 + 0 \cdot 2^4 + 0 \cdot 2^3$ ;
- $x_6 = 0 \cdot 2^7 + 1 \cdot 2^5 + 1 \cdot 2^4$ .
- The decimal equivalents of this p-adic representation of terminal nodes work out as  $x_1, x_2, \dots x_8 = 0, 2, 4, 32, 40, 48, 128, 192$ .
- A p-adic encoding for  $x_i$  is given by  $\sum_{1}^{n-1} a_k p_k$  where  $a_k \in \{0, 1\}$  and  $p_k = 2^k$ .

## **P-adic** (Algebraic) = Ultrametric (Topology)

- Various terms are used interchangeably for analysis in and over such fields such as p-adic, ultrametric, non-Archimedean, and isosceles.
- The natural geometric ordering of metric valuations is on the real line, whereas in the ultrametric case the natural ordering is a hierarchical tree.
- Ostrowski's theorem: Each non-trivial valuation on the field of the rational numbers is equivalent either to the absolute value function or to some p-adic valuation
- Alternatively: Up to equivalence, the only norms on the rationals are the p-adic norm and the usual norm given by the absolute value.

#### **Practical Interest of Ultrametricity**

- Hierarchies arise naturally in language syntax, and (it has been claimed) in financial markets.
- Rammal et al.: Ultrametricity is a natural property of high-dimensional spaces, and ultrametricity emerges as a consequence of randomness and of the law of large numbers.
- Again Rammal et al. and recent work of ours: Sparsely coded data tend to be ultrametric. Examples include: the use of complete disjunctive forms of coding in correspondence analysis; and categorical data coding in genomics and proteomics, speech, and other fields.
- Ultrametricity is considered to hold at low Planck scales, and in superstrings (Brekke and Freund, Phys. Rep., 233, 1–66, 1993).
- Also to be valid for optimization spaces.

#### **Testing for Ultrametricity**

- Rammal et al.: determine the subdominant ultrametric (aka single link hierarchic clustering).
- Interesting phase space effects for increase in dimensionality.
- However the subdominant ultrametric gives rise to pathologies.
- E.g. "friends of friends" chaining effect:  $d(x,y) \le r_0$ ,  $d(y,z) \le r_0$  then  $d(x,z) = 2r_0 \epsilon$  for arbitrarily small  $\epsilon$ . Hence d(x,z) can be anomalously large.

- A basic unifying framework for pairs of objects, and the distance valuation on them, is that of a *binary relation*.
- $\bullet$  On a set E, a binary relation is a *preorder* if it is reflexive and transitive;
- it is an *equivalence relation* if the binary relation is reflexive, transitive and symmetric;
- and it is an *order* if the binary relation is reflexive, transitive, and anti-symmetric.

• Let F denote the set of pairs of distinct units in E. A distance defines a total preorder on F:

$$\forall \{(x,y),(z,t)\} \in F : (x,y) \le (z,t) \Longleftrightarrow d(x,y) \le d(z,t)$$

• A preorder is called ultrametric if:

$$\forall x, y, z \in E : \rho(x, y) \le r \text{ and } \rho(y, z) \le r \Longrightarrow \rho(x, z) \le r$$

where r is a given integer and  $\rho(x,y)$  denotes the rank of pair (x,y) for  $\bar{\omega}$ .

• A necessary and sufficient condition for a distance on E to be ultrametric is that the associated preorder (on  $E \times E$ , or alternatively preordonnance on E) is ultrametric.

- We move on now to define Lerman's H-classifiability index, which measures how ultrametric a given metric is.
- Let M(x, y, z) be the median pair among  $\{(x, y), (y, z), (x, z)\}$  and let S(x, y, z) be the highest ranked pair among this triplet. J is the set of all such triplets of E.
- Mapping  $\tau$  of all triplets J into the open interval of all pairs F for the given preorder  $\omega$ :

$$\tau: J \longrightarrow ]M(x, y, z), S(x, y, z)[$$

- Given a triplet  $\{x,y,z\}$  for which  $(x,y) \leq (y,z) \leq (x,z)$ , for preorder  $\omega$ , the interval ]M(x,y,z), S(x,y,z)[ is empty if  $\omega$  is ultrametric. Relative to such a triplet, the preorder  $\omega$  is "less ultrametric" to the extent that the cardinal of ]M(x,y,z), S(x,y,z)[, defined on  $\omega$ , is large.
- $H(\omega) = \sum_{J} ||M(x, y, z), S(x, y, z)||/(|F| 3)|J||$

- Data sets that are "more classifiable" in an intuitive way, i.e. they contain "sporadic islands" of more dense regions of points a prime example is Fisher's iris data contrasted with 150 uniformly distributed values in  $\mathbb{R}^4$  such data sets have a smaller value of  $H(\omega)$ . For Fisher's data we find  $H(\omega) = 0.0899$ , whereas for 150 uniformly distributed points in a 4-dimensional hypercube, we find  $H(\omega) = 0.1835$ .
- Extensive tests carried out have shown that uniform data has values around 0.18 -0.21. Whereas with more sparsely coded data, etc., one finds values around 0.1 0.14.

- We took 3D cylanders defined by RA and Dec within a tight radius of a position, to limit the number of galaxies studied at any given time to around 500.
- We used data in (lower left block in Sloan data) low RA, near galactic plane.
- Then we used 3D uniformly distributed data to see how different the Lerman index would be.
- For Sloan data: 0.149837, 0.115096, 0.148676.
- For uniform data: 0.187662, 0.179590, 0.171903.
- Numbers in each case: 589, 554, 715.

#### **Conclusions and Critique**

- The Sloan data came out as more ultrametric in all cases, compared to uniformly distributed 3D values.
- But a Euclidean distance was used for determining the Lerman index.
- Also the cylandrical volume used in Sloan space may have biased the results (in view of the redshift value).
- Future work: replace the cylander with a cone, and study replacement for the Euclidean distance.