

Metric Curvature and Applications

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consider a number of applications of metric curvature to a variety of problems. Amongst them:
[A] The problem of better approximating surfaces by triangular meshes. We suggest to view the approximating triangulations (graphs) as finite metric spaces and the target smooth surface as

metric differential geometry for the analysis of weighted graphs/networks. In particular, Haantjes curvature as a tool in communication networks and DNA microarray analysis.

Embedding Curvature

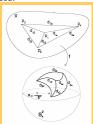
Our goal: to define an intrinsic metric curvature for sur-

We do this by comparing quadruples on the given metric space, to those in a gauge surface.

Definition Let (M,d) be a metric space, and let Q = $\{p_1,...,p_4\}\subset M$, together with the mutual distances: $d_{ij}=$ $a_{ii} = d(p_i, p_i); 1 \le i, j \le 4$. The set Q together with the set of distances $\{d_{ij}\}_{1 \leq i,j \leq 4}$ is called a *metric quadruple*.

Definition The *embedding curvature* $\kappa(Q)$ of the metric quadruple Q is defined be the curvature κ of S_{κ} into which can be isometrically embedded.





The Embedding Curvature at a point is defined by passing to the limit:

Definition Let (M,d) be a metric space, and let $p \in M$ be an accumulation point. Then \emph{p} is said to have \emph{Wald} curvature $\kappa_W(p)$ iff

- (i) No neighbourhood of p is linear;
- (ii) For any $\varepsilon>0$, exists $\delta>0$ s.t. (a) $Q=\{p_1,...,p_4\}\subset M$, and (b) $d(p, p_i) < \delta (i = 1, ..., 4) \Longrightarrow |\kappa(Q) - \kappa_W(p)| < \varepsilon$.

An approximate formula (Robinson):

Given the semi-dependent metric quadruple $Q = Q(p_1, p_2, p_3, p_4)$, of distances $d_{ij} = dist(p_i, p_j)$, i = 1, ..., 4, the embedding curvature $\kappa(Q)$ is well approximated by:

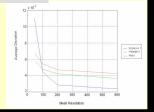
$$K(Q) = \frac{6(\cos \angle_0 2 + \cos \angle_0 2')}{d_{24}(d_{12}\sin^2(\angle_0 2) + d_{23}\sin^2(\angle_0 2'))};$$

where: $\angle_0 2 = \angle(p_1 p_2 p_4)$, $\angle_0 2' = \angle(p_3 p_2 p_4)$ represent the angles of the Euclidean triangles of sides d_{12}, d_{14}, d_{24} and d_{23}, d_{24}, d_{34} , respectively.



Experimental Results

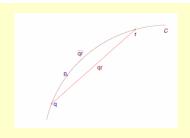




Haantjes Curvature

For Rectifiable Metric Spaces

Definition Let (M,d) be a metric space, let $c: I = [0,1] \stackrel{\sim}{\to} M$ be a homeomorphism, and let $p, q, r \in c(I), q, r \neq p$. Denote by \widehat{qr} the arc of c(I) between q and r, and by qr segment



Then c has *Haantjes Curvature* $\kappa_H(p)$ at the point p iff:

$$\kappa_H^2(p) = 24 \lim_{q,r \to p} \frac{l(\widehat{qr}) - d(q,r)}{\left(l(\widehat{qr})\right)^3}$$

where " $l(\widehat{qr})$ " denotes the length

Possible Application: As approximation of sectional curvature for triangulated surface reconstruction (see figure below).

For Weighted Graphs

Definition Let G = (V, G, d) be a metric graph, and let $v \in V$. Let $\pi = v_1 v v_2$ be a path through v. First we define the $\it curvature of triangles$ with vertex $\it v$ as being:

$$\kappa'_H(\triangle v_1vv_2) = \begin{cases} \sqrt{\frac{24^{|d(v_1,v)+d(v,v_2)-d(v_1,v_2)|}}{\left(d(v_1,v)+d(v,v_2)\right)^3}} & \epsilon = (v_1,v_2) \in E; \\ 0 & \epsilon = (v_1,v_2) \notin E. \end{cases}$$

Then the modified Haantjes curvature $\kappa'_{H,\pi} = \kappa'_H(v)$ of π at $\emph{\emph{v}}$ is defined to be the arithmetic mean of the curvatures of all the triangles with apex v:

$$\kappa_H'(p) = \frac{\sum\limits_{v_i \sim v, v_j \sim v, v_i \neq v_i} \kappa_H'(\triangle v_i v v_{ij})}{|\{\triangle v_i v v_{ij} | v_i \sim v, v_{ij} \sim v, v_{ij} \neq v_i\}|}$$

This represents a generalization of Combinatorial Curvature

Definition Let G be a (connected) graph and let v be a vertex of G, s.t. $ho(v) \geq 2$, where ho(v) denotes the degree of v i.e. $\rho(v) = |\{u \mid u \sim v\}|$. The combinatorial curvature of G at v is defined as:

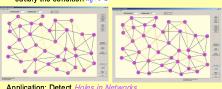
$$curv(v) \triangleq \frac{|\{\triangle vv_i v_{ij} | v_i \sim v, v_{ij} \sim v_i, v_{ij} \neq v\}|}{\rho(v)(\rho(v) - 1)/2}$$

that is, it represents the ratio between the actual number of triangles and the maximum number of possible triangles with apex at v.

Applications

Ouasi-Geodesics in Networks

se Haantjes curvature as *geodesic curvature* k_g in metric spaces, and define H-qu satisfy the condition k



Application: Detect



Communication Networks

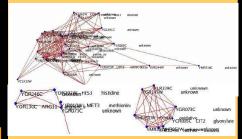


Clustering

To perform clustering, one selects a curvature threshold $T_{curv} \in [0,1]^*$ and selects a subgraph $H_{T_{curv}} \subseteq G$ by removing all nodes of curvature $< T_{curv}$ together with their

DNA microarray data taken from

http://rana.lbl.gov/EisenData.html is made into a graph by a method of correlation based "edging". Namely, one computes the correlation between different DNA microarrays and sets an edge between them according to a (correlation) threshold.



Here the metric is induced by the gene length, as they were shown to be relevant for the functioning of genes. More precisely we have employed - for the special case of gene length as weights - the following metric:

Definition Let (G, E, μ) be a connected vertex weighted graph. Define (for all $v \sim u$):

$$d(v,w) = \begin{cases} \frac{|\mu(v)| + |\mu(w)|}{|\mu(v)\mu(w)|} & v \neq w \text{, } \mu(v), \mu(w) \neq 0; \\ 0 & v \neq w \text{, } \mu(v) = 0 \text{ or } \mu(w) = 0; \\ 0 & v = w \text{.} \end{cases}$$