Aggregation on the Basis of Structure Recognition

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Structure Recognition

Structure recognition: identify specific cartographic objects or aggregates as well as spatial relations, and measures of importance

A city model consists mainly of buildings of differnet shape, style and roads.

Topics to be discussed

- Identification of objects (buildings).
- Identification of spatial relations.
- Aggregations rules based upon sturcture recogoniton.

Simple Buildings

Simple Buildings:







Gable



























Dutch hip

Complex buildings





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Gable – Lean

Arrow gable

b. building as a group of small buildings joined together forming a special shape



Closed square

bridge



cross



Arc

Trapezium



Building recognition using NN

NN-bsiac structure







Neural Netweok (NN) approach

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🗞 NeuNet Pro - NNShapeHeco - [Browse | est Hesults]

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Spatial relationship

Spatial relationships among individuals and neighborhood objects is based upon followings:

- i. Proximity
- ii. Containment
- iii. Connectivity

Closest

Contained

Connect

And can be divided into three levels:



Positional parameters





Other parameters

• Symmetry

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- Regularity
- Orthogonal walls (in most cases)
- Roof types
- General Shapes
- L & U junctions
- parallel lines
- parallelism of faces

Meso level:

- It applies to an object in relation to its neighbors. Out of the three relations described above, proximity is very important relations among 3D buildings in neighborhood.
- Proximity can me obtained by measuring their mutual orientation, angle, distance between them and their height difference



Macro level:

Applies to clusters of objects having similar properties such as settlement blocks and is based on visual grouping behaviors. These visual grouping behaviors are an important aspect in understanding images and maps and therefore has to be maintained. Macro level structure recognition helps to maintain this. Main emphasis is on

- Shape and size regularity.
- Height regularity.
- Regularity of roof structure and surface texture.
- Adjacency to other structures.
- Unique, deterministic features.
- Relative distributional density

Aggregation rules

Aggregation rules are based upon structure recogniton and can be divided into following groups as follow:

Linkage rules define the spatial relations between buildings that must exist for their aggregation.

- Proximity: Two buildings must be disjoint but within a certain distance of each other for their aggregation;
- -Alignment: Two buildings should be aligned or their alignment should not differ much and the difference must be below the permissible limits.
- Angle: Two buildings should have minimum angle between them if they are to be aggregated;
- Height: Two buildings should have minimum height difference if they are to be aggregated. Roof Type: Two buildings should have similar type of roof, i.e., planar or gable if they are to be aggregated;
- Adjacency: Two buildings must have their adjacent faces close if they are to be aggregated

Semantic rules: define the semantic relationships that must exist for aggregation. Semantic rules include relationships such as

- Class: Two buildings must be of the same class
- Structural: a group of objects form a common geographic or perceptual structure.

Orientational rules define the historical or local importance of the buildings. Important buildings such as monument, theatre, tower etc. reflect the city identity.

- Buildings which are of great importance should have their identities preserved at maximum of its extent. Therefore two buildings having historical importance should not be aggregated.
- If only one of the buildings is important, even then aggregation should not be done. Instead the important building should be exaggerated. The same rule applies to other local important buildings like TV towers, theatres, schools etc.

Based upon the above structural recognition, an algorithm for aggregation was developed. It is basically a nested set of If..Else statements. For example:

If $d(o_i, o_j) < \Delta d_{min}$ then AND If $h(o_i, o_j) < \Delta h_{min}$ then AND If $roof_type_1 = roof_type_2$ then AND If $\Delta A(o_i, o_j) < A_{min}$ then aggregate









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Future work:

Macro level spatial relation based upon visual grouping will be studied and consequently quantified to judge the quality of generalization. These relation are based upon:

- Parallelism
- Continuity
- Closure
- **Proximity**
- Orientation

Thanks!

