

Assessment of Quality of Vectorization Based Information

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The presented paper discusses results of the experiment on an assessment of quality of information, derived from raster to vector conversion. Contents of the test object are disclosed. Three test object of different qualitative content are discussed. An assessment of information quality was carried out on a basis of a comparison with initial information with information obtained after vectorization. Coefficients are introduced which describe quality of information after processing.

Key words: Image processing, image scanning, Information vectorization,
Assessment of quality of information

During image processing^{1,2} vectorization, which is a procedure of a conversion of raster information into vector information, is of importance³. It is used in design⁴, geo-informatics, where it serves as a basis for a formation of Geodata⁵. In practice, specialized software for vectorization⁶ is implemented, which solves the major part of the problem of information conversion. That software use special filters⁷, which set up the process of recognition for a certain type of information⁸, such as drawings, text, business graphics, geographical maps, etc. The vectorization is a process of a transformation of information⁹ from one form to another. In practice a semantic component of the process plays an important role¹⁰. The result of information processing using such software is not always unambiguous. Furthermore, because of the specific characteristics of an algorithm, it is possible to

identify some vectorization's systematic errors. Special testing allows to identify and to take into account such errors. Special test objects are used for that.

Methodology

In order to test vectorization's quality, three test objects were used. The test object is an object, which semantic and metric contents are known prior to processing. Such an object allows to compare processing result with initial information and to make a conclusion about processing quality basing on that. An equivalent of test objects are bench marks^{11, 12}, which are used for testing of software's performance. The test object includes a set of primitive object which define statistics. Using that statistics, it is possible to conduct statistical estimations characterizing processing quality.

The first test object was an aggregate of primitive objects for recognition, which were forming a regular grid of 100 crosses with 10mm vertical distance and 1.5 mm horizontal distance between them. The second test object was an aggregate of 100 horizontal segments with 1.5 mm

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size. The third test object was an aggregate of 80 vertical segments with 1.5 mm size.

The following parameters were used: N – number of input objects before processing; N_i – number of distorted interpreted objects; N_{ni} – number of lost objects; N_{on} – number of output non distorted objects; N_{ni} – number of distorted not interpreted objects;

Following coefficients were introduced K_l – coefficient of losses; K_{inf} – coefficient of informativeness; K_d – coefficient of distortion; K_{int} – coefficient of interpretability of distorted objects; – coefficient of representativeness.

$$K_l = N_i / N; K_{inf} = (N_i + N_o) / N; K_d = (N_i + N_{ni}) / N; K_{int} = N_i / (N_i + N_{ni}); = N_o / N.$$

The object was considered as a distorted, if there was a change of number of elements in it and if its center of symmetry shifted relatively to an initial position.

RESULTS

Software scan was carried out in a vertical direction. Sweepline algorithm was implemented^{13, 14}. The results are presented in the table.

Table 1. Vectorization results in Vectory 3.1 software with three test objects

| | Test 1 100 | Test 2 100 | Test 3 80 |
|-----------|---------------|---------------|--------------|
| N_i | 11 | 14 | 0 |
| N_{ni} | 6 | 0 | 7 |
| N_{ni} | 6 | 8 | 8 |
| N_o | 77 | 72 | 65 |
| κ | 0.77 | 0.72 | 0.81 |
| K_l | 0.11 | 0.14 | 0.0 |
| K_d | 0.12 | 0.08 | 0.23 |
| K_{inf} | 0.83 | 0.72 | 0.9 |
| K_{int} | 0.5 | 0.0 | 0.47 |

DISCUSSION

Coefficient of representativeness is the main characteristic of quality of processed information. It shows, which part of objects remains its semantic and metric properties.

That procedure for quality assessment of information is applicable for test objects with

different number of recognizable objects. It is applicable for an assessment of information quality not only software, but also scanning devices.

The results, presented in the table, allow to presume, that objects' scan direction affects quality of information. The results presented the table suggest, that there are spatial relationships between scan direction and scan quality of converted information. The results are of interest not only for engineering applications, but also for theoretical research in a field of spatial knowledge. Informativeness is determined not on a basis of data volume, but by a number of interpreted objects after processing. This is semantic approach.

CONCLUSION

An expansion of types of recognition objects in a form of other shapes, such as circles, is of certain interest. Moreover, recognition objects, which were formed as crossing of primitive objects or their tangential junction, are also interesting. Is is reasonable to expand research area to spatial knowledge^{15, 16}, which is investigated using artificial intelligence methods¹⁷.

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