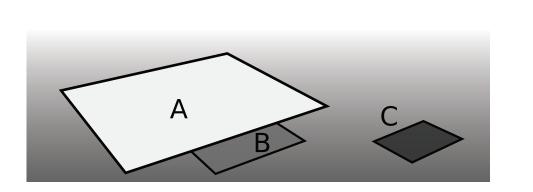
Using extrusion to generate higher-dimensional GIS datasets

Ken Arroyo Ohori* and Hugo Ledoux Delft University of Technology, The Netherlands

TUDelft

1. Introduction



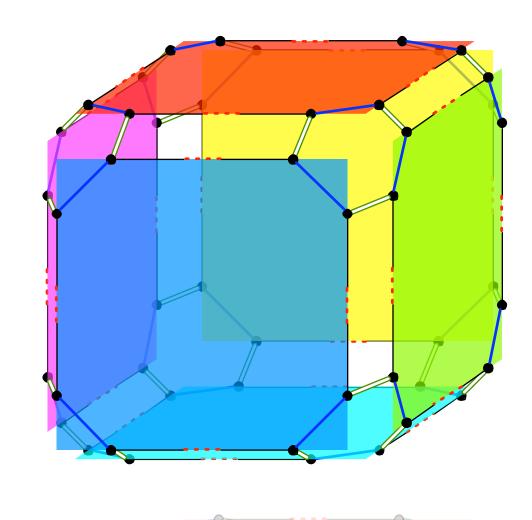
There is a growing interest in the use of higher-dimensional (>4D) digital objects, but their use is hampered by a lack of methods and algorithms.

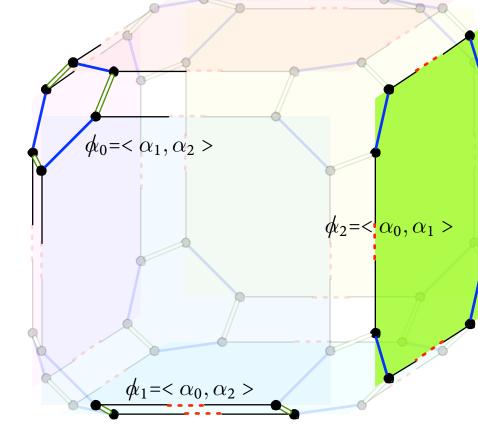
Akin to 2D to 3D extrusion (left), nD extrusion allows us to create an (n + 1)-dimensional model from an n-dimensional one by assigning it a range along the (n+1)-th dimension.

We present here a dimensionindependent extrusion algorithm for linear geometries using generalised maps. It is optimal and straightforward to implement.

We have implemented it in Python and made experiments in which 2D real-world GIS datasets were extruded to 3D and 4D.

2. Generalised maps





A simple implementation (right) is based on structures per dart and per geometric embedding.

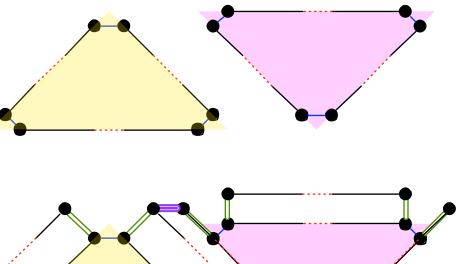
Generalised maps (G-maps) are a topological model in arbitrary dimensions developed by Lienhardt (1994). It is composed of darts and involutions (left). A dart is a combination of a cell of every dimension. Involutions connect darts related along a certain dimension.

```
struct Dart {
    Dart *involutions[n+1];
    Embeddings *embeddings[n+1];
};

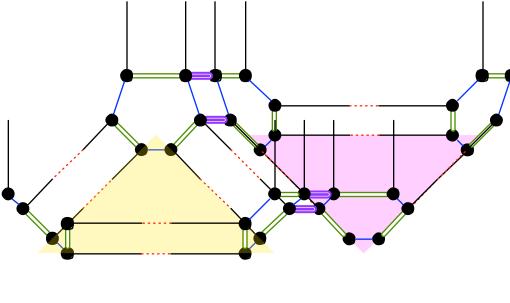
struct Embedding {
    Dart *referenceDart;
    Embedding *holes[];
    int dimension;
    ...
    float red, green, blue;
};

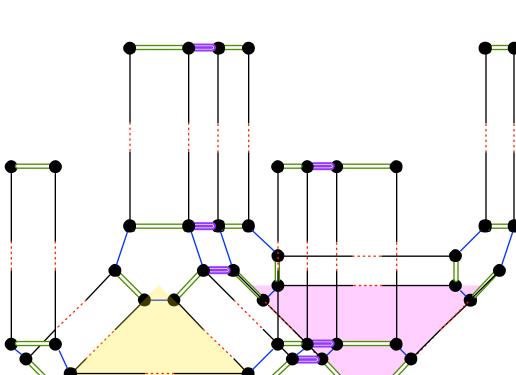
struct PointEmbedding : Embedding {
    float x, y, z;
}
```

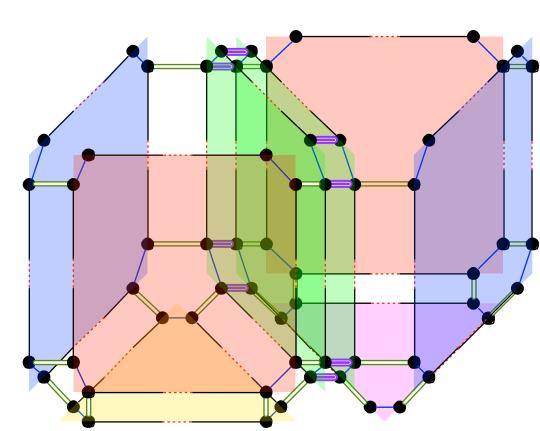
3. Extrusion algorithm

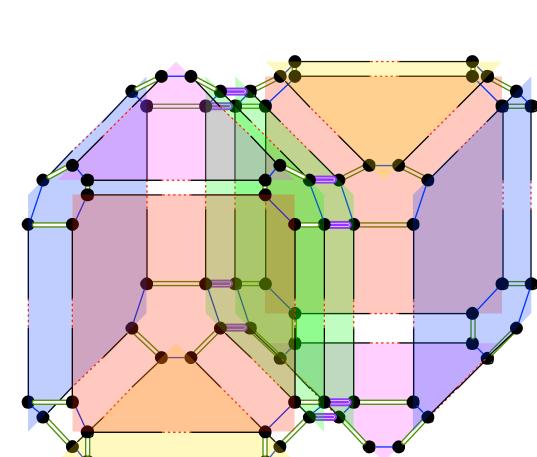


Our extrusion algorithm takes two input arguments: an (n-1)-G-map, and a given range $[r_{\min}, r_{\max}]$ where these will exist along an n-th dimension. Its result is an n-G-map representing a set of prismatic n-polytopes.









Algorithm 1: EmbeddingsExtrusion

Input : E, $[r_{\min}, r_{\max}]$ Output: E', base, top, exforeach $e \in E$ do $base(e), top(e), ex(e) \leftarrow e$ if e.dimension = 0 then $Append r_{\min} \text{ to } base(e)\text{'s coordinates}$ $Append r_{\max} \text{ to } top(e)\text{'s coordinates}$ $ex(e).dimension \leftarrow ex(e).dimension + 1$ Put base(e), top(e) and ex(e) in E'

Algorithm 2: GMAPEXTRUSION

```
Input : G = (D, \alpha_0, \alpha_1, \dots, \alpha_{n-1}), E, E', base, top, ex, e, e'
Output: G' = (D', \alpha'_0, \alpha'_1, \dots, \alpha'_n)
for dim \leftarrow n to 0 do
\begin{bmatrix} GMAPLAYER(G, G', dim, 1, last, E, E', e, e', base, ex) \\ last \leftarrow cur \end{bmatrix}
for dim \leftarrow 0 to n do
\begin{bmatrix} GMAPLAYER(G, G', dim, 0, last, E, E', e, e', top, ex) \\ last \leftarrow cur \end{bmatrix}
```

Algorithm 3: GMAPLAYER

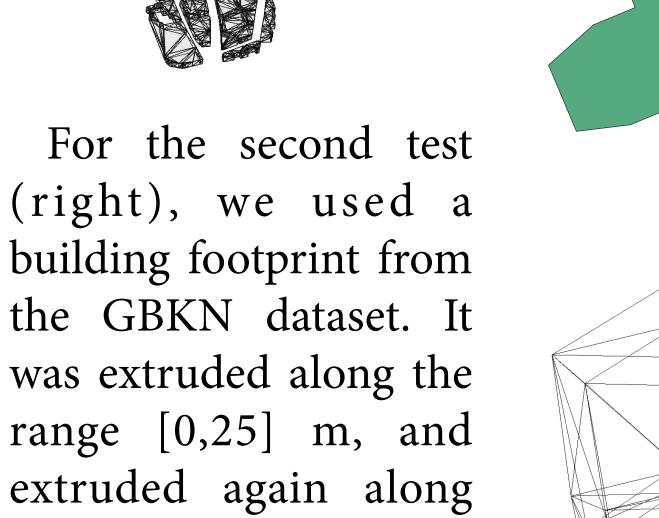
```
Input : G = (D, \alpha_0, \alpha_1, \dots, \alpha_{n-1}), G' = (D', \alpha'_0, \alpha'_1, \dots, \alpha'_n),
          dim, offset, last, E, E', e, e', el, ex
Output: G' = (D', \alpha'_0, \alpha'_1, \dots, \alpha'_n), last, cur, e'
foreach d \in D do
      cur(d) \leftarrow \text{new dart}
      Put cur(d) in D'
foreach d \in D do
      for inv \leftarrow 0 to dim - 1 do
            \alpha'_{inv}(cur(d)) \leftarrow cur(\alpha_{inv}(d))
      \alpha'_{dim+offset}(cur(d)) \leftarrow last(d)
      \alpha'_{dim+offset}(last(d)) \leftarrow cur(d)
      for inv \leftarrow dim + 2 to n do
            \alpha'_{inv}(cur(d)) \leftarrow cur(\alpha_{inv-1}(d))
      for emb \leftarrow 0 to dim do
            e'_{emb}(cur(d)) \leftarrow el(e_{emb}(d))
      for emb \leftarrow dim + 1 to n do
            e'_{emb}(cur(d)) \leftarrow ex(e_{emb-1}(d))
```

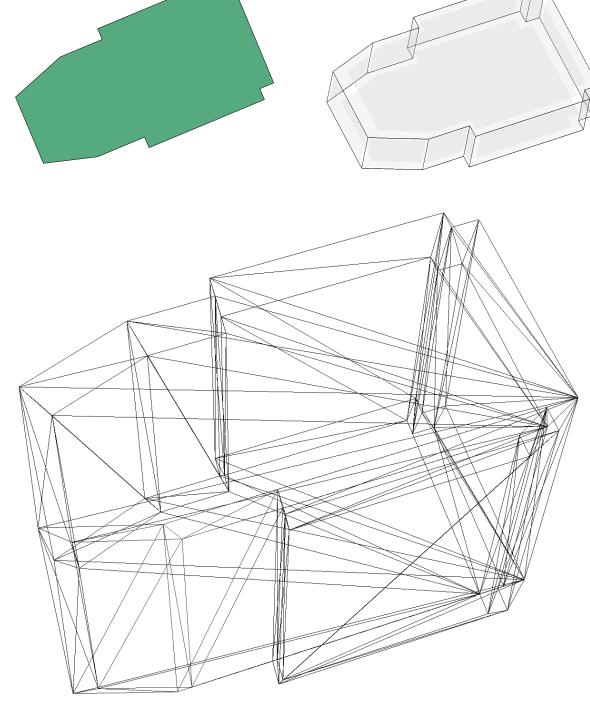
4. Experiments

We have implemented the algorithm in Python and tested it with a few datasets in the area of Delft.

For the first test (left), an area of the TOPIONI dataset

For the first test (left), an area of the TOP10NL dataset was extruded along the range [0,15] m. We verified the results using the procedure described in Ledoux and Meijers (2011), based on a constrained tetrahedralisation of the dataset.





5. Discussion

We have shown that it is possible to use our algorithm to extrude (n-1)-dimensional cell complexes represented as G-maps into n-dimensional ones. Our algorithm is optimal in time, easy to implement and addresses both the geometry and the topology of the objects.

References

Pascal Lienhardt. N-dimensional generalized combinatorial maps and cellular quasi-manifolds. *International Journal of Computational Geometry and Applications*, 4(3):275-324, 1994.

Hugo Ledoux and Martijn Meijers. Topologically consistent 3D city models obtained by extrusion. *International Journal of Geographical Information Science*, 25(4):557-574, 2011.

Contact

tudelft.nl/kenohorig.a.k.arroyoohori@tudelft.nl

the range [1960,2060]

yr for a 4D (3D+time)

model.

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