The basic larcc module *

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1 Basic representations

A few basic representation of topology are used in LARCC. They include some common sparse matrix representations: CSR (Compressed Sparse Row), CSC (Compressed Sparse Column), COO (Coordinate Representation), and BRC (Binary Row Compressed).

1.1 BRC (Binary Row Compressed)

We denote as BRC (Binary Row Compressed) the standard input representation of our LARCC framework. A BRC representation is an array of arrays of integers, with no requirement of equal length for the component arrays. The BRC format is used to represent a (normally sparse) binary matrix. Each component array corresponds to a matrix row, and contains the indices of columns that store a 1 value. No storage is used for 0 values.

BRC format example Let $A = (a_{i,j} \in \{0,1\})$ be a binary matrix. The notation BRC(A) is used for the corresponding data structure.

$$A = \begin{pmatrix} 0,1,0,0,0,0,0,1,0,0 \\ 0,0,1,0,0,0,0,0,0,0 \\ 1,0,0,1,0,0,0,0,0,1 \\ 1,0,0,0,0,0,1,1,1,0,0 \\ 0,0,1,0,1,0,0,0,1,0 \\ 0,0,0,0,0,0,0,0,0,0,0 \\ 0,1,0,0,0,0,0,0,0,0,0 \\ 0,1,1,0,0,0,0,0,0,0,0 \\ 0,1,1,0,1,0,0,0,0,0,0 \end{pmatrix} \mapsto \begin{array}{c} [[1,7], \\ [2], \\ [0,3,9], \\ [0,6], \\ [2,4,8], \\ [2], \\ [2,4,8], \\ [1,7,9], \\ [3,8], \\ [1,2,4]] \end{array}$$

1.2 Format conversions

Macro referenced in 49b.

From triples to scipy.sparse The function brc2Coo transforms a BRC representation in a list of triples (row, column, 1) ordered by row.

Two coordinate compressed sparse matrices cooFV and cooEV are created below, starting from the BRC representation FV and EV of the incidence of vertices on faces and edges, respectively, for a very simple plane triangulation.

```
⟨Test example of Brc to Coo transformation 3a⟩ ≡
    print "\n>>> brc2Coo"
    V = [[0, 0], [1, 0], [2, 0], [0, 1], [1, 1], [2, 1]]
    FV = [[0, 1, 3], [1, 2, 4], [1, 3, 4], [2, 4, 5]]
    EV = [[0,1],[0,3],[1,2],[1,3],[1,4],[2,4],[2,5],[3,4],[4,5]]
    cooFV = brc2Coo(FV)
    cooEV = brc2Coo(EV)
    assert cooFV == [[0,0,1],[0,1,1],[0,3,1],[1,1,1],[1,2,1],[1,4,1],[2,1,1],
    [2,3,1], [2,4,1],[3,2,1],[3,4,1],[3,5,1]]
    assert cooEV == [[0,0,1],[0,1,1],[1,0,1],[1,3,1],[2,1,1],[2,2,1],[3,1,1],
    [3,3,1],[4,1,1],[4,4,1],[5,2,1],[5,4,1],[6,2,1],[6,5,1],[7,3,1],[7,4,1],
    [8,4,1],[8,5,1]]
    ◊
```

Macro referenced in 50.

Conversion to csr format Then we give the function triples2mat to make the transformation from the sparse matrix, given as a list of triples row, column, value (non-zero elements), to the scipy.sparse format corresponding to the shape parameter, set by default to "csr", that stands for Compressed Sparse Row, the normal matrix format of the LARCC framework.

```
⟨From list of triples to scipy.sparse 3b⟩ ≡

def triples2mat(triples,shape="csr"):
    n = len(triples)
    data = arange(n)
    ij = arange(2*n).reshape(2,n)
    for k,item in enumerate(triples):
        ij[0][k],ij[1][k],data[k] = item
    return scipy.sparse.coo_matrix((data, ij)).asformat(shape)
```

Macro referenced in 49b.

The conversion from triples to csr format is provided below.

Macro referenced in 49b.

Two CSR sparse matrices csrFV and csrEV are generated (by *scipy.sparse*) in the following example:

 \langle Test example of Coo to Csr transformation 3d $\rangle \equiv$

```
csrFV = coo2Csr(cooFV)
csrEV = coo2Csr(cooEV)
print "\ncsr(FV) =\n", repr(csrFV)
print "\ncsr(EV) =\n", repr(csrEV)
```

Macro referenced in 50.

The *scipy* printout of the last two lines above is the following:

Conversion from BRC to CSR format The transformation from BRC to CSR format is implemented slightly differently, according to the fact that the matrix dimension is either unknown (shape=(0,0)) or known.

```
⟨ Brc to Csr transformation 4a⟩ ≡
  def csrCreate(BRCmatrix,lenV=0,shape=(0,0)):
    triples = brc2Coo(BRCmatrix)
  if shape == (0,0):
        CSRmatrix = coo2Csr(triples)
  else:
        CSRmatrix = scipy.sparse.csr_matrix(shape)
        for i,j,v in triples: CSRmatrix[i,j] = v
    return CSRmatrix
```

Macro referenced in 49b.

Example The conversion to CSR format of the characteristic matrix faces-vertices FV is given below for our simple example made by four triangle of a manifold 2D space, graphically shown in Figure 1a. The LAR representation with CSR matrices does not make difference between manifolds and non-manifolds, conversely than most modern solid modelling representation schemes, as shown by removing from FV the third triangle, giving the model in Figure 1b.

```
⟨Test example of Brc to Csr transformation 4b⟩ ≡
    print "\n>>> brc2Csr"
    V = [[0, 0], [1, 0], [2, 0], [0, 1], [1, 1], [2, 1]]
    FV = [[0, 1, 3], [1, 2, 4], [1, 3, 4], [2, 4, 5]]
    EV = [[0,1], [0,3], [1,2], [1,3], [1,4], [2,4], [2,5], [3,4], [4,5]]
    csrFV = csrCreate(FV)
    csrEV = csrCreate(EV)
    print "\ncsrCreate(FV) =\n", csrFV
```

Macro referenced in 7, 50.

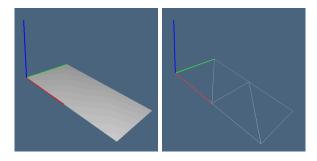


Figure 1: (a) Simplicial 2-complex; (b) its 1-skeleton.

2 Matrix operations

As we know, the LAR representation of topology is based on CSR representation of sparse binary (and integer) matrices. In this section we hence discuss the stack of matrix representations and operations implemented by this module. The current python prototype makes reference to the scipy implementation of sparse matrices. Later implementations in different languages will necessarily make reference to different matrix packages.

2.1 Basic operations

Two utility functions allow to query the number of rows and columns of a CSR matrix, independently from the low-level implementation (that in the following is provided by scipy.sparse).

Sparse to dense matrix transformation The Scipy package provides the useful method .todense() in order to transform any sparse matrix format in the corresponding dense format. The function csr2DenseMatrix is given here for the sake of generality and portability.

Matrix product and transposition The following macro provides the IDE interface for the two main matrix operations required by LARCC, the binary product of compatible matrices and the unary transposition of matrices.

```
⟨ Matrix product and transposition 6c ⟩ ≡
    def matrixProduct(CSRm1,CSRm2):
        CSRm = CSRm1 * CSRm2
        return CSRm

def csrTranspose(CSRm):
        CSRm = CSRm.T
        return CSRm

⋄
Macro referenced in 49b.
```

2.2 Characteristic matrices

We define as characteristic matrices M_k ($0 \le k \le d$) the binary matrices having as rows the images of the characteristic functions of the k-cells $\alpha_k \subset V$ of a cellular complex with vertices V. Remember that characteristic (or *indicator*) function is

$$\mathbf{1}_A : V \to \{0, 1\},\$$

which for a given subset A of X, has value 1 at points of A and 0 at points of V-A.

Example: from BRC to CSR to dense matrix Let us compute and show in dense form the characteristic matrices of 2- and 1-cells of the simple manifold given in Figure 1. By running the file test/py/larcc/test08.py the reader will get the two matrices shown in Example 2

```
"test/py/larcc/test08.py" 7 ≡

""" Characteristic matrices """

from larlib import *

⟨Test example of Brc to Csr transformation 4b⟩

⟨Test examples of Sparse to dense matrix transformation 6b⟩

⋄
```

Example 1 (Dense Characteristic matrices). Let us notice that the two matrices below have the some numbers of columns (indexed by vertices of the cell decomposition). This very fact allows to multiply one matrix for the other transposed, and hence to compute the matrix form of linear operators between the spaces of cells of various dimensions.

$$FV = \begin{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ & \begin{bmatrix} 1 & 1 & 0 & 1 & 0 & 0 \\ & \begin{bmatrix} 0 & 1 & 1 & 0 & 1 & 0 \\ & \begin{bmatrix} 0 & 1 & 1 & 0 & 1 & 0 \\ & \begin{bmatrix} 0 & 1 & 0 & 1 & 1 & 0 \end{bmatrix} \end{bmatrix} \end{bmatrix}$$

$$EV = \begin{bmatrix} \begin{bmatrix} 1 & 1 & 0 & 0 & 0 & 0 \\ & \begin{bmatrix} 1 & 0 & 0 & 1 & 0 & 0 \\ & \begin{bmatrix} 0 & 1 & 1 & 0 & 0 & 1 \\ & \begin{bmatrix} 0 & 1 & 0 & 0 & 1 & 0 \\ & \begin{bmatrix} 0 & 0 & 1 & 0 & 1 & 0 \\ & \begin{bmatrix} 0 & 0 & 1 & 1 & 0 \end{bmatrix} \end{bmatrix} \end{bmatrix}$$

$$\begin{bmatrix} \begin{bmatrix} 0 & 0 & 1 & 0 & 1 & 1 \\ & \begin{bmatrix} 0 & 0 & 0 & 1 & 1 \\ & \begin{bmatrix} 0 & 0 & 0 & 1 & 1 \end{bmatrix} \end{bmatrix}$$

Example 2 (Operators from edges to faces and vice-versa). As a general rule for operators between two spaces of chains of different dimensions supported by the same cellular complex, we use names made by two characters, whose first letter correspond to the target space, and whose second letter to the domain space. Hence FE must be read as the operator from edges to faces. Of course, since this use correspond to see the first letter as the space generated by rows, and the second letter as the space generated by columns. Notice that the element

(i,j) of such matrices stores the number of vertices shared between the (row-)cell i and the (column-)cell j.

$$FE = FV EV^{\top} = \begin{bmatrix} \begin{bmatrix} 2 & 2 & 1 & 2 & 1 & 0 & 0 & 1 & 0 \\ & \begin{bmatrix} 1 & 0 & 2 & 1 & 2 & 2 & 1 & 1 & 1 \\ & \begin{bmatrix} 1 & 1 & 1 & 2 & 2 & 1 & 0 & 2 & 1 \\ & \begin{bmatrix} 0 & 0 & 1 & 0 & 1 & 2 & 2 & 1 & 2 \end{bmatrix} \end{bmatrix}$$

$$EF = EV FV^{\top} = \begin{bmatrix} \begin{bmatrix} 2 & 1 & 1 & 0 \\ & 2 & 0 & 1 & 0 \\ & & \begin{bmatrix} 1 & 2 & 1 & 1 \\ & 2 & 1 & 2 & 0 \end{bmatrix} \\ & \begin{bmatrix} 0 & 2 & 1 & 2 \\ & 0 & 1 & 0 & 2 \end{bmatrix} \\ & \begin{bmatrix} 0 & 1 & 0 & 2 \\ & 1 & 2 & 1 \end{bmatrix} \\ & \begin{bmatrix} 0 & 1 & 1 & 2 \end{bmatrix} \end{bmatrix}$$



Figure 2: example caption

Matrix elements filtering Some filtering operations on matrix elements are needed in the implementation of various topological operators. Some of such filtering operations are given below.

```
⟨ Matrix filtering to produce the boundary matrix 8⟩ ≡

def csrBoundaryFilter(CSRm, facetLengths):
    maxs = [max(CSRm[k].data) for k in range(CSRm.shape[0])]
    inputShape = CSRm.shape
    coo = CSRm.tocoo()
    for k in range(len(coo.data)):
        if coo.data[k] ==maxs[coo.row[k]]: coo.data[k] = 1
        else: coo.data[k] = 0

    mtx = coo_matrix((coo.data, (coo.row, coo.col)), shape=inputShape)
    out = mtx.tocsr()
```

```
return out
Macro referenced in 49b.
\langle Test example of Matrix filtering to produce the boundary matrix 9a \rangle \equiv
     print "\n>>> csrBoundaryFilter"
     csrEF = matrixProduct(csrFV, csrTranspose(csrEV)).T
     facetLengths = [csrCell.getnnz() for csrCell in csrEV]
     CSRm = csrBoundaryFilter(csrEF, facetLengths).T
     print "\ncsrMaxFilter(csrFE) =\n", csr2DenseMatrix(CSRm)
Macro referenced in 50.
\langle Matrix filtering via a generic predicate 9b \rangle \equiv
     def csrPredFilter(CSRm, pred):
          # can be done in parallel (by rows)
          coo = CSRm.tocoo()
          triples = [[row,col,val] for row,col,val
                       in zip(coo.row,coo.col,coo.data) if pred(val)]
          i, j, data = TRANS(triples)
          CSRm = scipy.sparse.coo_matrix((data,(i,j)),CSRm.shape).tocsr()
          return CSRm
Macro referenced in 49b.
\langle Test example of Matrix filtering via a generic predicate 9c\rangle \equiv
     print "\n>>> csrPredFilter"
     CSRm = csrPredFilter(matrixProduct(csrFV, csrTranspose(csrEV)).T, GE(2)).T
     print "\nccsrPredFilter(csrFE) =\n", csr2DenseMatrix(CSRm)
Macro referenced in 50.
```

Correction to boundary operator The more deep issue with LAR (that already appeared at the beginning of LAR experiments, two years ago) is discussed and solved here.

The boundary operator (and the whole LAR approach) works well with convex cells, but MAY give errors (like $\partial \partial \neq 0$) with non-convex cells. At that time, we understood the source of the problem, due to the fact that ALL the vertices of a (d-1)-cell EXTERNAL to a NON-CONVEX d-cell may belong to its boundary ... (e.g. BOTH the vertices of a 1-cell external to a non-convex 2-cell)

You might say: convexity is a geometric concept, boundary is topological, so what? Because the input is certainly geometrical, via drawings and/or machine generated EV,FV starting from drawings, so the two aspects are strongly interconnected ...

We implement here a general resolution strategy, since it is pretty easy to discover the (d-1)-facets where the problem appears: they are shared by MORE than TWO d-cells ... and this one is a topological ERROR. So, testing $\partial \partial$ on each face, we eliminate the problem within the boundary operator, but not within the LAR data structure (to be marked in some way where the problem may appear ... TODO).

My guess is that this source of errors might appear anytime the Boolean input becomes pretty complex, and that it was the greatest source of random errors.

```
"test/py/larcc/test20.py" 10 \equiv
     """ Correction to boundary operator """
     from larlib import *
     V,[VV,EV,FV] = larCuboids([2,1],True)
     complex = Struct([(V,FV,EV), t(.5,.25), s(.5,.5), (V,FV,EV)])
     V,FV,EV = struct2lar(complex)
     lines = [[V[v] for v in edge] for edge in EV]
     V,FV,EV,polygons = larFromLines(lines)
     VV = AA(LIST)(range(len(V)))
     submodel = STRUCT(MKPOLS((V,EV)))
     VIEW(larModelNumbering(1,1,1)(V,[VV,EV,FV],submodel,.5))
     csrBoundaryMat = boundary(FV,EV)
     print "wrong boundary matrix =",csrBoundaryMat.todense()
     csrBoundaryMat = boundary1(FV,EV,VV) # <<<< NOTE !!</pre>
     print "right boundary matrix =",csrBoundaryMat.todense()
     \langle 2-boundary example 11, ... \rangle
     submodel = viewBoundaryChain((V,FV,EV))([1,1,1,1])
     VIEW(larModelNumbering(1,1,1)(V,[VV,EV,FV],submodel,.5))
     submodel = viewBoundaryChain((V,FV,EV))([0,0,1,1])
     VIEW(larModelNumbering(1,1,1)(V,[VV,EV,FV],submodel,.5))
```

The printing from example test/py/larcc/test20.py is given below. Notice the difference between the fourth lines from the bottom of the two matrices, generated by the function applications boundary(FV,EV) and boundary1(FV,EV,VV), respectively.

```
wrong boundary matrix = [
                                          right boundary matrix = [
 [0 0 1 0]
                                            [0 0 1 0]
 [0 0 1 1]
                                            [0 0 1 1]
 [0 1 1 0]
                                            [0 1 1 0]
 [0 1 1 0]
                                            [0 1 1 0]
 [0 0 1 0]
                                            [0 0 1 0]
 [0 0 0 1]
                                            [0 0 0 1]
 [1 0 0 1]
                                            [1 0 0 1]
 [0 0 0 1]
                                            [0 0 0 1]
 [1 0 0 1]
                                            [1 0 0 1]
 [0 0 1 0]
                                            [0 0 1 0]
 [0 0 0 1]
                                            [0 0 0 1]
 [1 1 1 1]
                                            [0 0 0 0]
 [1 0 0 1]
                                            [1 0 0 1]
 [0 1 1 0]
                                            [0 1 1 0]
 [0 0 1 1]]
                                            [0 0 1 1]]
```

Figure 3: (a) A 2-complex with non-convex cells; (b) boundary of 2-chain [1,1,1,1]; (c) boundary of 2-chain [0,0,1,1].

Whereas the boundary 2-chain is

```
⟨2-boundary example 11⟩ ≡
    """ View boundary chain """
    def viewBoundaryChain(larModel):
        V,FV,EV = larModel
        VV = AA(LIST)(range(len(V)))
        def viewBoundaryChain0 (chain):
            BE = chain2BoundaryChain(boundary1(FV,EV,VV))(chain)
            if chain == [1,1,1,1]: assert BE == [0,4,5,7,9,10]
            if chain == [0,0,1,1]: assert BE == [0,2,3,4,5,6,7,8,9,10,12,13]
        BEV = [EV[e] for e in BE]
        if chain == [1,1,1,1]: assert BEV == [(0,1),(5,6),(7,8),(5,7),(0,6),(1,8)]
        submodel = STRUCT(MKPOLS((V,BEV)))
        return submodel
        return viewBoundaryChain0
```

Macro defined by 11, 12a. Macro referenced in 10.

The cellular 2-complex generated by example The printing from example test/py/larcc/test20.py is displayed in Figure ??. The corresponding LAR is the following:

```
 \begin{array}{l} \langle \text{ 2-boundary example } 12a \rangle \equiv \\ & \quad \forall = [[0.0,1.0],[1.0,1.0],[1.0,0.75],[1.0,0.25],[0.5,0.25],[1.0,0.0],\\ & \quad [0.0,0.0],[2.0,0.0],[2.0,1.0],[0.5,0.75],[1.5,0.25],[1.5,0.75]] \\ & \quad \text{FV} = [[3,2,11,10],[9,2,3,4],[1,2,9,4,3,5,6,0],[1,8,7,5,3,10,11,2]] \\ & \quad \text{EV} = [(0,1),(1,2),(4,9),(2,9),(5,6),(7,8),(3,10),(5,7),(2,11),(0,6),\\ & \quad (1,8),(2,3),(10,11),(3,4),(3,5)] \\ & \quad \Diamond \\ & \quad \text{Macro defined by } 11,12a. \\ & \quad \text{Macro referenced in } 10. \\ \end{array}
```

2.3 Boundary operator on 3-complex with non-convex cells

Correction to boundary operator

```
"test/py/larcc/test21.py" 12b \equiv
     """ Correction to boundary operator """
     from larlib import *
     V = [[0.0,0.0,0.0],[0.0,0.0,1.0],[0.0,1.0,0.0],[0.0,1.0,1.0],[1.0,0.0,0.0],
     [1.0,0.0,1.0],[1.0,1.0,0.0],[1.0,1.0,1.0],[2.0,0.0,0.0],[2.0,0.0,1.0],
     [2.0,1.0,0.0], [2.0,1.0,1.0], [0.5,0.25,0.25], [0.5,0.25,0.75], [0.5,0.75,
     0.25, [0.5,0.75,0.75], [1.0,0.25,0.25], [1.0,0.25,0.75], [1.0,0.75,0.25],
     [1.0,0.75,0.75], [1.5,0.25,0.25], [1.5,0.25,0.75], [1.5,0.75,0.25], [1.5,0.75,0.25]
     0.75, 0.75]]
     CV = [(16,17,18,19,20,21,22,23),(4,5,6,7,8,9,10,11,16,17,18,19,20,21,
     22,23), (12,13,14,15,16,17,18,19), (0,1,2,3,4,5,6,7,12,13,14,15,16,
     17,18,19)]
     FV = [(0,2,4,6),(12,13,14,15),(4,6,8,10),(0,1,4,5),(16,17,18,19),(4,5,6,
     7,16,17,18,19),(17,19,21,23),(16,17,20,21),(0,1,2,3),(18,19,22,23),
     (14,15,18,19),(1,3,5,7),(12,13,16,17),(12,14,16,18),(13,15,17,19),
     (4,5,8,9),(2,3,6,7),(16,18,20,22),(5,7,9,11),(6,7,10,11),(8,9,10,11),
     (20,21,22,23)
     EV = [[0,1],[2,3],[4,5],[6,7],[8,9],[10,11],[0,2],[1,3],[4,6],[5,7],
     [8,10], [9,11], [0,4], [1,5], [2,6], [3,7], [4,8], [5,9], [6,10], [7,11], [12,13],
     [14,15], [16,17], [18,19], [20,21], [22,23], [12,14], [13,15], [16,18], [17,19],
     [20,22],[21,23],[12,16],[13,17],[14,18],[15,19],[16,20],[17,21],[18,22],
     [19,23]
```

```
csrBoundaryMat = boundary(CV,FV)
print "wrong boundary matrix =",csrBoundaryMat.todense()
csrBoundaryMat = boundary1(CV,FV,EV)
print "right boundary matrix =",csrBoundaryMat.todense()

VV = AA(LIST)(range(len(V)))
submodel = STRUCT(MKPOLS((V,EV)))
VIEW(larModelNumbering(1,1,1)(V,[VV,EV,FV,CV],submodel,.5)))
```

Testing correction to boundary operator for general (non-convex) LAR The test example exported in test/py/larcc/test22.py is aimed to check the working of LAR on chain complexes supported by cellular decompositions with non-convex but connected cells, possibly of topological genus > 1, i.e. with more than one hole in some face.

```
"test/py/larcc/test22.py" 13a ≡

""" Testing correction to boundary operator for general (non-convex) LAR """

from larlib import *

⟨Input and visualization of a general cellular complex 13b⟩
⟨Visualization of the skeletons 15a⟩
⟨Computation of corrected boundary operator 15b⟩
⟨Visualization of some boundary chains 16⟩

⋄
```

Input and visualization of a general cellular complex

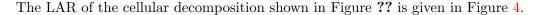
```
(Input and visualization of a general cellular complex 13b) ≡
    """ Input and visualization of a general cellular complex """

mod1 = larQuote1D([0.2,-0.2,0.2])
    mod2 = larQuote1D([0.5,0.5])
    mod3 = larQuote1D([0.5])
    mod = larModelProduct([larModelProduct([mod2,mod1]),mod3]))

mx = larQuote1D([1,1])
    my = larQuote1D([1])
    m = larModelProduct([larModelProduct([mx,my]),my]))

complex = Struct([m, t(.5,.2,.25), mod])
    V,CVO = struct2lar(complex)
    CV = copy(CVO)
    CV[1] += CV[0] + CV[4]
    CV[3] += CV[2] + CV[5]
    VIEW(SKEL_1(STRUCT( MKPOLS((V,CV))))))
```

Macro referenced in 13a.



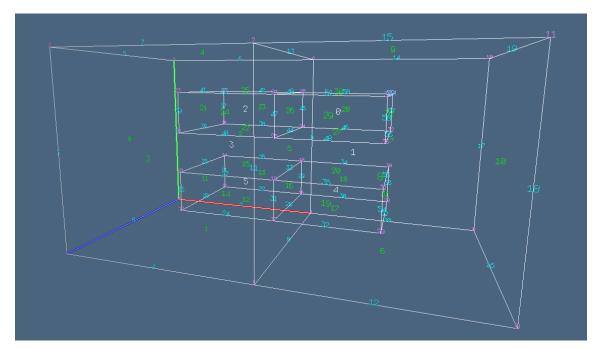


Figure 4: The cellular 3-complex defined below. Cells of different dimension are numbered in different colours.

We have 36 vertices, indexed from 0 to 35, and six 3-cells, numbered from 0 to 5. Notice that the subsets of integers CV[1] and CV[3] provide the description of the two non-convex cells.

```
 V = [[0.0,0.0,0.0], [0.0,0.0,1.0], [0.0,1.0,0.0], [0.0,1.0,1.0], [1.0,0.0,0.0], \\ [1.0,0.0,1.0], [1.0,1.0,0.0], [1.0,1.0,1.0], [2.0,0.0,0.0], [2.0,0.0,1.0], \\ [2.0,1.0,0.0], [2.0,1.0,1.0], [0.5,0.2,0.25], [0.5,0.2,0.75], [0.5,0.4,0.25], \\ [0.5,0.4,0.75], [1.0,0.2,0.25], [1.0,0.2,0.75], [1.0,0.4,0.25], [1.0,0.4,0.25], \\ [0.5,0.6,0.25], [0.5,0.6,0.75], [0.5,0.8,0.25], [0.5,0.8,0.75], [1.0,0.6,0.25], \\ [1.0,0.6,0.75], [1.0,0.8,0.25], [1.0,0.8,0.75], [1.5,0.2,0.25], [1.5,0.2,0.75], \\ [1.5,0.4,0.25], [1.5,0.4,0.75], [1.5,0.6,0.25], [1.5,0.6,0.75], [1.5,0.8,0.25], \\ [1.5,0.8,0.75]]
```

```
CV = [(24,25,26,27,32,33,34,35),(4,5,6,7,8,9,10,11,24,25,26,27,32,33,34,35,16,17,18,19,28,29,30,31),(20,21,22,23,24,25,26,27),(0,1,2,3,4,5,6,7,20,21,22,23,24,25,26,27,12,13,14,15,16,17,18,19),(16,17,18,19,28,29,30,31),(12,13,14,15,16,17,18,19)]
```

Computation and visualization of the skeletons

```
\langle Visualization of the skeletons 15a \rangle \equiv
     """ Visualization of the skeletons """
     def cubeFV(verts):
         [a,b,c,d, e,f,g,h] = verts
         return [[a,b,c,d],[e,f,g,h],[a,c,e,g],[b,d,f,h],[a,b,e,f],[c,d,g,h]]
     FV = COMP([sorted,set,AA(tuple),CAT,AA(cubeFV)])(CVO)
     V,EV = larFacets((V,FV),dim=2)
     VV = AA(LIST)(range(len(V)))
     submodel = STRUCT(MKPOLS((V,EV)))
     VIEW(larModelNumbering(1,1,1)(V,[VV,EV,FV,CV],submodel,.25))
     FV[5] += FV[16] + FV[26]
     assert 30==len(boundaryCells(CVO,FV)) # assembly of 3 unrelated 3-complexes
Macro referenced in 13a.
FV = [(0,1,2,3),(0,1,4,5),(0,2,4,6),(1,3,5,7),(2,3,6,7),
(4,5,6,7,16,17,18,19,24,25,26,27),
(4,5,8,9),(4,6,8,10),(5,7,9,11),(6,7,10,11),(8,9,10,11),(12,13,14,15),
(12,13,16,17),(12,14,16,18),(13,15,17,19),(14,15,18,19),(16,17,18,19),
(16,17,28,29),(16,18,28,30),(17,19,29,31),(18,19,30,31),(20,21,22,23),
(20,21,24,25), (20,22,24,26), (21,23,25,27), (22,23,26,27), (24,25,26,27),
(24,25,32,33),(24,26,32,34),(25,27,33,35),(26,27,34,35),(28,29,30,31),
(32,33,34,35)
```

Computation of corrected boundary operator Of course, totalChain = len(CV)*[1] is the binary coordinate [1,1,1,1,1] of the cellular chain made by all the six 3-cells in CV. Analogously, BF is the list of integer indices of boundary 2-faces, corresponding to the coordinate representation given below, i.e. to the binary string of length 33, equating the size of the 2-skeleton of the cellular complex.

Conversely, BEV is (a superset of) the 1-skeleton cells of the boundary, according to the image in Figure ??.

```
\langle Computation of corrected boundary operator 15b\rangle \equiv """ Computation of corrected boundary operator """ totalChain = len(CV)*[1]
```

```
BF = chain2BoundaryChain(boundary1(CV,FV,EV))(totalChain)
BFV = [FV[f] for f in BF]
VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,BFV))))
glass = MATERIAL([1,0,0,0.1, 0,1,0,0.1, 0,0,1,0.1, 0,0,0,0.1, 100])
VIEW(STRUCT(AA(glass)(MKPOLS((V,BFV)))))
VIEW(glass(EXPLODE(1.2,1.2,1.2)(MKPOLS((V,BFV)))))
V,BEV = larFacets((V,BFV),dim=1)
VIEW(EXPLODE(1.2,1.2,1.2)(MKPOLS((V,[e for e in BEV if len(e)==2]))))
```

Macro referenced in 13a.

Visualization of some boundary chains

```
\( \text{Visualization of some boundary chains 16} \) \( = \)
\( \text{""" Visualization of some boundary chains """} \)

\( \text{BF} = \text{chain2BoundaryChain(boundary1(CV,FV,EV))([1,0,1,1,0,0])} \)
\( \text{BFV} = [FV[f] \) \for f in BF] \\
\( \text{V,BEV} = \text{larFacets((V,BFV),dim=1)} \)
\( \text{BEV} = [e \text{for e in BEV if len(e)==2}] \\
\( \text{VIEW(glass(STRUCT(MKPOLS((V,BFV))))} \)
\( \text{submodel} = \text{STRUCT(MKPOLS((V,BEV)))} \)
\( \text{VIEW(larModelNumbering(1,1,1)(V,[VV,BEV,BFV],submodel,.25))} \)
\( \text{obj} = \text{Struct([(V,BFV,BEV)])} \)
\( \text{VIEW(EXPLODE(1.2,1.2,1.2)(MKPOLS((W,EW))))} \)
\( \text{VIEW(EXPLODE(1.2,1.2,1.2)(MKPOLS((W,EW))))} \)
\( \text{VIEW(SKEL_1(EXPLODE(1.2,1.2,1.2)(MKTRIANGLES((W,FW,EW)))))} \)
\( \text{Application of some boundary chains """
\( \text{BFV} = \text{Constant of the property of the prope
```

Macro referenced in 13a.

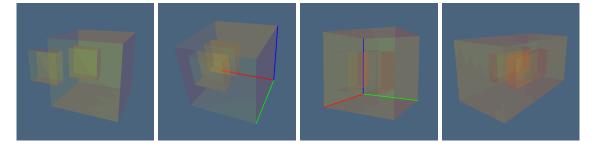


Figure 5: example caption

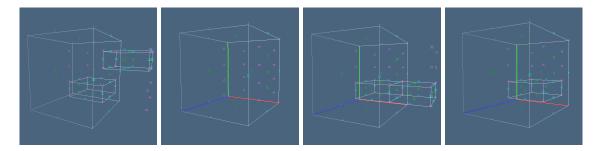


Figure 6: example caption

Relational inversion (characteristic matrix transposition) The operation could be executed by simple matrix transposition of the CSR (Compressed Sparse Row) representation of the sparse characteristic matrix $M_d \equiv \text{CV}$. A simple relational inversion using Python lists is given here. The invertRelation function is given here, linear in the size of the CV list, where the complexity of each cell is constant and small in most cases.

```
⟨Characteristic matrix transposition 17⟩ ≡
    """ Characteristic matrix transposition """
    def invertRelation(CV):
        def myMax(List):
            if List==[]: return -1
             else: return max(List)

        columnNumber = max(AA(myMax)(CV))+1
        VC = [[] for k in range(columnNumber)]
        for k,cell in enumerate(CV):
            for v in cell: VC[v] += [k]
        return VC
```

Macro referenced in 49b.

2.4 Computation of lower-dimensional skeletons

In most cases, in particular when the cellular complex is made by convex cells, the only cells of maximal dimension must be entered to gain a complete knowledge of the whole complex. Here we show how to compute the (d-1)-skeleton of a complex starting from its d-dimensional skeleton.

Extraction of facets of a cell complex The following larFacets function returns the LAR model V,cellFacets starting from the input model parameter. Two optional parameters define the (intrinsic) dimension of the input cells, with default value equal to three, and the eventual presence of a emptyCellNumber of empty cells. Their number

default to zero when the complex is closed, for example in the case it provides the d-boundary of a (d+1)-complex. If empty cells are present, their subset must be located at the end of the cell list.

```
\langle Extraction of facets of a cell complex 18a\rangle \equiv
     def setup(model,dim):
         V, cells = model
         csr = csrCreate(cells)
         csrAdjSquareMat = larCellAdjacencies(csr)
         csrAdjSquareMat = csrPredFilter(csrAdjSquareMat, GE(dim)) # ? HOWTODO ?
         return V,cells,csr,csrAdjSquareMat
     def larFacets(model, dim=3, emptyCellNumber=0):
         """ Estraction of (d-1)-cellFacets from "model" := (V,d-cells)
              Return (V, (d-1)-cellFacets)
         V,cells,csr,csrAdjSquareMat = setup(model,dim)
         solidCellNumber = len(cells) - emptyCellNumber
         cellFacets = []
         # for each input cell i
         for i in range(len(cells)):
              adjCells = csrAdjSquareMat[i].tocoo()
              cell1 = csr[i].tocoo().col
              pairs = zip(adjCells.col,adjCells.data)
              for j,v in pairs:
                  if (i<j) and (i<solidCellNumber):</pre>
                      cell2 = csr[j].tocoo().col
                      cell = list(set(cell1).intersection(cell2))
                      cellFacets.append(sorted(cell))
         # sort and remove duplicates
         cellFacets = sorted(AA(list)(set(AA(tuple)(cellFacets))))
         return V, cellFacets
Macro referenced in 49b.
\langle Computation of cell adjacencies 18b\rangle \equiv
     def larCellAdjacencies(CSRm):
         CSRm = matrixProduct(CSRm,csrTranspose(CSRm))
         return CSRm
Macro referenced in 49b.
```

Examples Two simple complexes are defined below by providing the pair V,FV. In both cases the EV relation is computed via the larFacets function.

```
\langle Test examples of Extraction of facets of a cell complex 19a\rangle \equiv
     """ A first (simplicial) example """
     V = [[0.,0.],[3.,0.],[0.,3.],[3.,3.],[1.,2.],[2.,2.],[1.,1.],[2.,1.]]
     FV = [[0,1,3],[1,2,4],[2,4,5],[3,4,6],[4,6,7],[5,7,8], # full
          [1,3,4],[4,5,7], # empty
         [0,1,2],[6,7,8],[0,3,6],[2,5,8] # exterior
     _,EV = larFacets((V,FV),dim=2)
     print "\nEV =",EV
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,EV))))
     """ Another (cuboidal) example """
     FV = [[0,1,6,7],[0,2,4,6],[4,5,6,7],[1,3,5,7],[2,3,4,5],[0,1,2,3]]
     _,EV = larFacets((V,FV),dim=2)
     print "\nEV =",EV
     VV = AA(LIST)(range(len(V)))
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,EV))))
Macro referenced in 50.
```

Visualization of cell numbers The adjacentcy matrices between 2-cells and 1-cells are printed here. Finally, the complex is displayed by numbering with different colours and sizes (depending on the rank) the complex cells.

```
Test examples of Computation of cell adjacencies 19b =

print "\n>>> larCellAdjacencies"
adj_2_cells = larCellAdjacencies(csrCreate(FV))
print "\nadj_2_cells =\n", csr2DenseMatrix(adj_2_cells)
adj_1_cells = larCellAdjacencies(csrCreate(EV))
print "\nadj_1_cells =\n", csr2DenseMatrix(adj_1_cells)

submodel = mkSignedEdges((V,EV))
VIEW(submodel)
VIEW(larModelNumbering(scalx=1,scaly=1,scalz=1)(V,[VV,EV,FV],submodel,2))
```

Macro referenced in 50.

3 Topological operations

In this section we provide the matrix representation of operators to compute the more important and useful topological operations on cellular complexes, and/or the indexed relations they return. We start the section by giving a graphical tool used to test the developed software, concerning the graphical writing of the full set of indices of the cells of every dimension in a 3D cuboidal complex.

3.1 Visualization of cellular complexes

It is often necessary to have a visual picture of the generated structures and computations. This section provides some quite versatile visualisation tools of both the cells and/or their integer indices.

Visualization of cell indices As already outlined, the modelIndexing function return the hpc value assembling both the 1-skeletons of the cells of every dimensions, and the graphical output of their indices, located on the centroid of each cell, and displayed using colors and sizes depending on the rank of the cell.

```
\langle Visualization of cell indices 20a \rangle \equiv
     """ Visualization of cell indices """
     from larlib import *
     def modelIndexing(shape):
         V, bases = larCuboids(shape,True)
         # bases = [[cell for cell in cellComplex if len(cell) == 2**k] for k in range(4)]
         color = [ORANGE, CYAN, GREEN, WHITE]
         nums = AA(range)(AA(len)(bases))
         hpcs = []
         for k in range(4):
              hpcs += [SKEL_1(STRUCT(MKPOLS((V,bases[k]))))]
              hpcs += [cellNumbering((V,bases[k]),hpcs[2*k])(nums[k],color[k],0.3+0.2*k)]
         return STRUCT(hpcs)
Macro defined by 20ab.
Macro referenced in 49b.
\langle Visualization of cell indices 20b \rangle \equiv
     """ Numbered visualization of a LAR model """
     def larModelNumbering(scalx=1,scaly=1,scalz=1):
         def larModelNumberingO(V,bases,submodel,numberScaling=1):
              color = [ORANGE, CYAN, GREEN, WHITE]
              nums = AA(range)(AA(len)(bases))
              hpcs = [submodel]
              for k in range(len(bases)):
                  hpcs += [cellNumbering((V,bases[k]),submodel)
                                (nums[k],color[k],(0.5+0.1*k)*numberScaling)]
              return STRUCT(hpcs)
              #return EXPLODE(scalx,scaly,scalz)(hpcs)
         return larModelNumbering0
     \Diamond
Macro defined by 20ab.
Macro referenced in 49b.
```

Drawing of oriented edges The following function return the hpc of the drawing with arrows of the oriented 1-cells of a 2D cellular complex. Of course, each edge orientation is from second to first vertex, independently from the vertex indices. Therefore, the edge orientation can be reversed by swapping the vertex indices in the 1-cell definition.

```
\langle Drawing of oriented edges 21a\rangle \equiv
     """ Drawing of oriented edges (2D) """
     def mkSignedEdges (model,scalingFactor=1):
         V, EV = model
         assert len(V[0])==2
         hpcs = []
         times = C(SCALARVECTPROD)
         frac = 0.06*scalingFactor
         for e0,e1 in EV:
             v0,v1 = V[e0], V[e1]
             vx,vy = DIFF([v1, v0])
             nx,ny = [-vy, vx]
             v2 = SUM([v0, times(0.66)([vx,vy])])
             v3 = SUM([v0, times(0.6-frac)([vx,vy]), times(frac)([nx,ny])])
             v4 = SUM([v0, times(0.6-frac)([vx,vy]), times(-frac)([nx,ny])])
             verts, cells = [v0,v1,v2,v3,v4],[[1,2],[3,4],[3,5]]
             hpcs += [MKPOL([verts,cells,None])]
         hpc = STRUCT(hpcs)
         return hpc
```

Example of oriented edge drawing An example of drawing of oriented edges is given in test/py/larcc/test11.py file, and in Figure 7, showing both the numbering of the cells and the arrows indicating the edge orientation is illustrated in Figure 7, where also the oriented boundary is shown.

Macro referenced in 49b.

```
"test/py/larcc/test11.py" 21b \(\text{Example of oriented edge drawing 21c}\) \(\phi\)
\(\text{Example of oriented edge drawing 21c}\) \(\text{Example of oriented edge drawing 21c}\) \(\text{Example of oriented edge drawing """}\)
\(\text{from larlib import *}\)
\(\text{V} = [[9,0],[13,2],[15,4],[17,8],[14,9],[13,10],[11,11],[9,10],[7,9],[5,9],[3,8],[0,6],[2,3],[2,1],[5,0],[7,1],[4,2],[12,10],[6,3],[8,3],[3,5],[5,5],[7,6],[8,5],[10,5],[11,4],[10,2],[13,4],[14,6],[13,7],[11,9],[9,7],[7,7],[4,7],[2,6],[12,7],[12,5]]
```

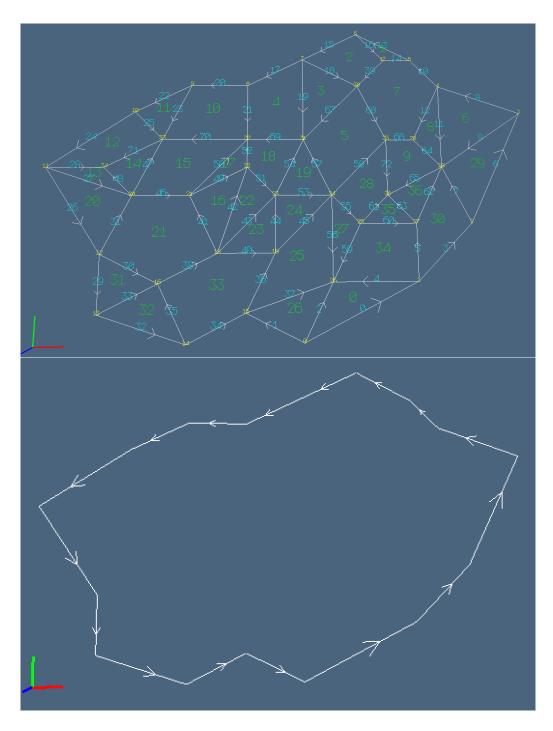


Figure 7: Example of numbered polytopal complex, including edge orientations, and its oriented boundary.

```
FV = [[0,1,26],[5,6,17],[6,7,17,30],[7,30,31],[7,8,31,32],[24,30,31,35],[3,4,
28],[4,5,17,29,30,35],[4,28,29],[28,29,35,36],[8,9,32,33],[9,10,33],[11,10,
33,34],[11,20,34],[20,33,34],[20,21,32,33],[18,21,22],[21,22,32],[22,23,31,
32],[23,24,31],[11,12,20],[12,16,18,20,21],[18,22,23],[18,19,23],[19,23,24],
[15,19,24,26],[0,15,26],[24,25,26],[24,25,35,36],[2,3,28],[1,2,27,28],[12,13,
16],[13,14,16],[14,15,16,18,19],[1,25,26,27],[25,27,36],[36,27,28]]
VIEW(EXPLODE(1.2,1.2,1)(MKPOLS((V,FV))))
VV = AA(LIST)(range(len(V)))
_,EV = larFacets((V,FV+[range(16)]),dim=2,emptyCellNumber=1)
submodel = mkSignedEdges((V,EV))
VIEW(submodel)
VIEW(larModelNumbering(scalx=1,scaly=1,scalz=1)(V,[VV,EV,FV],submodel,2))
orientedBoundary = signedCellularBoundaryCells(V,[VV,EV,FV])
cells = [EV[e] if sign==1 else REVERSE(EV[e]) for (sign,e) in zip(*orientedBoundary)]
submodel = mkSignedEdges((V,cells))
VIEW(submodel)
```

Extracting the boundary of whichever chain The boundary of whichever chain, here defined as the list of indices of its cells, then transformed to its coordinate representation (column vector in the given basis), is explicitly computed by matrix product times the matrix of the boundary operator in the given basis, transformed back in its BRC representation, and displayed as LAR model.

```
"test/py/larcc/test19.py" 23 \=
    """ Example of oriented edge drawing """

    \langle Example of oriented edge drawing 21c \rangle

    C2 = csr_matrix((len(FV),1))
    for i in [1,2, 12,13,14,15, 22,23, 29,30,31]: C2[i,0] = 1

    BD = boundary(FV,EV)
    C1 = BD * C2
    C_1 = [i for i in range(len(EV)) if ABS(C1[i,0]) == 1 ]
    C_2 = [i for i in range(len(FV)) if C2[i,0] == 1 ]

    VIEW(EXPLODE(1.2,1.2,1)(MKPOLS((V,[EV[k] for k in C_1] + [FV[k] for k in C_2]))))
    \langle
```

From 2D chains to boundary chains

Macro referenced in 21b, 23.

```
\langle From 2D chains to boundary chains 24\rangle \equiv
     """ From 2D chains to boundary chains """
     def chain2BoundaryChain(csrBoundaryMat):
         print ">>> chain2BoundaryChain"
         nedges,nfaces = csrBoundaryMat.shape
         def chain2BoundaryChain0(chain):
             row = np.array(range(len(chain)))
              col = np.array([0 for k in range(len(chain))])
              data = np.array(chain)
             csrFaceVect = scipy.sparse.coo_matrix((data, (row, col)), shape=(nfaces,1)).tocsr()
             csrEdgeVect = csrBoundaryMat*csrFaceVect
             boundaryChain = [h for h,val in
                  zip(csrEdgeVect.tocoo().row, csrEdgeVect.tocoo().data) if val%2 != 0]
              return boundaryChain
         print "<<< chain2BoundaryChain"</pre>
         return chain2BoundaryChain0
```

Macro referenced in 49b.

3.2 Incidence and adjacency operators

Let us start by computing the more interesting subset of the binary relationships between the 4 decompositive and/or boundary entities of 3D cellular models. Therefore, in this case we denote with C, F, E, and V, the 3-cells and their faces, edges and vertices, respectively. The input is the full-fledged LAR representation provided by

$$CV := CSR(M_3) \tag{1}$$

$$FV := CSR(M_2) \tag{2}$$

$$EV := CSR(M_1) \tag{3}$$

$$VV := CSR(M_0) \tag{4}$$

Of course, $CSR(M_0)$ coincides with the identity matrix of dimension |V| and can by excluded by further considerations. Some binary incidence and adjacency relations we are going to compute are:

$$CF := CV \times FV^t = CSR(M_3) \times CSR(M_2)^t$$
(5)

$$CE := CV \times EV^t = CSR(M_3) \times CSR(M_1)^t$$
(6)

$$FE := FV \times EV^t = CSR(M_2) \times CSR(M_1)^t \tag{7}$$

The other possible operators follow from a similer computational pattern.

The programming pattern for incidence computation A high-level function larIncidence useful to compute the LAR representation of the incidence matrix (operator) and the incidence relations is given in the script below.

```
⟨Some incidence operators 25a⟩ ≡
    """ Some incidence operators """
    def larIncidence(cells,facets):
        csrCellFacet = csrCellFaceIncidence(cells,facets)
        cooCellFacet = csrCellFacet.tocoo()
        larCellFacet = [[] for cell in range(len(cells))]
        for i,j,val in zip(cooCellFacet.row,cooCellFacet.col,cooCellFacet.data):
            if val == 1: larCellFacet[i] += [j]
        return larCellFacet

        ⟨Cell-Face incidence operator 25b⟩
        ⟨Cell-Edge incidence operator 25c⟩
        ⟨Face-Edge incidence operator 26a⟩
        ◇

Macro referenced in 49b.
```

Cell-Face incidence The csrCellFaceIncidence and larCellFace functions are given below, and exported to the larce module.

```
⟨ Cell-Face incidence operator 25b⟩ ≡
    """ Cell-Face incidence operator """
    def csrCellFaceIncidence(CV,FV):
        return boundary(FV,CV)

def larCellFace(CV,FV):
        return larIncidence(CV,FV)
        ◊

Macro referenced in 25a.
```

Cell-Edge incidence Analogously, the csrCellEdgeIncidence and larCellFace functions are given in the following script.

```
⟨ Cell-Edge incidence operator 25c⟩ ≡
    """ Cell-Edge incidence operator """
    def csrCellEdgeIncidence(CV,EV):
        return boundary(EV,CV)

def larCellEdge(CV,EV):
    return larIncidence(CV,EV)
```

Macro referenced in 25a.

Face-Edge incidence Finally, the csrCellEdgeIncidence and larCellFace functions are provided below.

Example The example below concerns a 3D cuboidal grid, by computing a full LAR stack of bases CV, FV, EV, VV, showing its fully numbered 3D model, and finally by computing some more useful binary relationships (CF, CE, FE), needed for example to compute the signed matrices of boundary operators.

```
"test/py/larcc/test10.py" 26b =
    """ A mesh model and various incidence operators """
    from larlib import *
    shape = [2,2,2]
    V,(VV,EV,FV,CV) = larCuboids(shape,True)
    VIEW(modelIndexing(shape))

CF = larCellFace(CV,FV)
    CE = larCellFace(CV,EV)
    FE = larCellFace(FV,EV)
```

3.2.1 Incidence chain

Let denote with CF, FE, EV the three consecutive incidence relations between k-cells and (k-1)-cells $(3 \le k \le 0)$ in a 3-complex. In the general multidimensional case, let us call CF_d the generic binary incidence operator, between d-cells and (d-1)-facets, as:

$$CF_d = M_{d-1}M_d^t,$$

with

$$\mathrm{CF}_d := \{a_{ij}\}, \qquad a_{ij} = \left\{ \begin{array}{ll} 1 & \text{if } M_{d-1}(i)M_d(j) = |f_j| \\ 0 & \text{otherwise} \end{array} \right.$$

Incidence chain computation The function incidenceChain, given below, returns the full stack of BRC incidence matrices of a LAR representation for a cellular complex, starting from its list of bases, i.e. from [VV,EV,FV,CV,...]. Notice that the function returns the inverse sequence [EV,FE,CF,...], i.e., CF_k $(1 \le k \le d)$.

```
\langle Incidence chain computation 27a\rangle \equiv
     """ Incidence chain computation """
     def incidenceChain(bases):
         #print "\n len(bases) = ",len(bases),"\n"
         pairsOfBases = zip(bases[1:],bases[:-1])
         relations = [larIncidence(cells,facets)
                          for cells,facets in pairsOfBases]
         return REVERSE(relations)
Macro referenced in 49b.
"test/py/larcc/test13.py" 27b \equiv
     """ Example of incidence chain computation """
     from larlib import *
     shape = (1,1,2)
     print "\n\nFor a better example provide a greater shape!"
     V,bases = larCuboids(shape,True)
     VV,EV,FV,CV = bases
     incidence = incidenceChain([VV,EV,FV,CV])
     relations = ["CF","FE","EV"]
     for k in range(3):
         print "\n\n incidence", relations[k], "=\n", incidence[k],
     print "\n\n"
     submodel = SKEL_1(STRUCT(MKPOLS((V,EV))))
     VIEW(larModelNumbering(scalx=1,scaly=1,scalz=1)(V,[VV,EV,FV,CV],submodel,1))
```

Example of incidence chain computation When running the test/py/larcc/test13.py file one obtains the following printout. Notice that it provides the links between d-cell numerations and the numerations of their faces. See Figure 8 for this purpose.

```
\langle \text{Incidence chain for a 3D cuboidal complex } 27c \rangle \equiv \\ \text{incidence CF} = [[0,2,4,6,8,9],[1,3,5,7,9,10]] \\ \text{incidence FE} = [[0,2,8,9],[1,3,9,10],[4,6,11,12],[5,7,12,13],[0,4,14,15], \\ [1,5,15,16],[2,6,17,18],[3,7,18,19],[8,11,14,17],[9,12,15,18],[10,13,16,19]]
```

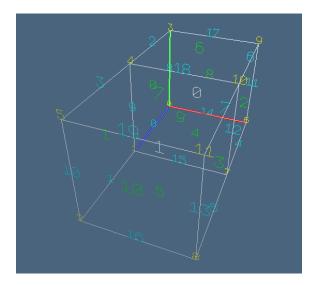


Figure 8: Che stack of incidence relations gives the common links between cell numerations.

```
incidence EV = [[0,1],[1,2],[3,4],[4,5],[6,7],[7,8],[9,10],[10,11],[0,3], [1,4],[2,5],[6,9],[7,10],[8,11],[0,6],[1,7],[2,8],[3,9],[4,10],[5,11]]
```

Macro never referenced.

3.3 Boundary and coboundary operators

When computing the matrices of boundary and coboundary operators it may be useful to distinguish between simplicial complexes and general polytopal complexes, including cuboidal ones. In the first cases all skeletons, and hence the other topological operators, may be computed using only combinatorial methods. In the second case some reference to their geometric embedding must be done, at least to compute the *oriented* boundary and coboundary. Therefore we separate the two cases in the following sections.

Computation of topological relations The function crossRelation is used here to compute a topological relation starting from two characteristic matrices XV and YV, that associate the sets of topological objects X and Y with their vertices, respectively. The technique using sparse binary matrices stored in CSR (Compressed Sparse Row) format is used.

```
⟨ Computation of topological relation 28⟩ ≡
    """ Computation of topological relation """
    def crossRelation(lenV,XV,YV,terminal=False):
        if len(YV) == len(CAT(YV)) == lenV:
```

```
return XV
else:
    csrXV = csrCreate(XV,lenV)
    csrYV = csrCreate(YV,lenV)
    csrXY = matrixProduct(csrXV, csrYV.T)
    XY = [None for k in range(len(XV))]
    for k,face in enumerate(XV):
        data = csrXY[k].data
        col = csrXY[k].indices
        XY[k] = [col[h] for h,val in enumerate(data) if val==min(len(XV[k]),len(YV[h]))]
        # NOTE: val depends on the relation under consideration ...
    return XY
```

Macro referenced in 49b.

3.3.1 Non-oriented operators

The boundary function below takes as parameters the BRC representations of d-cells and (d-1)-facets, and returns the CSR matrix of the boundary operator. Let us notice that such operator uses a mod 2 algebra, since it takes elements within the field $\mathbb{Z}_2 = \{0,1\}$.

Macro referenced in 50.

In the script below it is necessary to guarantee that both csrFV and csrCV are created with the same number of column. The initial steps have this purpose.

```
⟨ From cells and facets to boundary operator 29b⟩ ≡
    from boundary import boundary
    from boundary import boundary1

def coboundary(cells,facets):
```

```
Boundary = boundary(cells,facets)
         return csrTranspose(Boundary)
     def coboundary1(cells,facets):
         Boundary = boundary(cells,facets)
         return csrTranspose(Boundary)
Macro referenced in 49b.
\langle From cells and facets to boundary cells 30a\rangle \equiv
     def totalChain(cells):
         return csr_matrix(len(cells)*[[1]])
     def boundaryCells(cells,facets):
         csrBoundaryMat = boundary.boundary(cells,facets)
         csrChain = csr_matrix(totalChain(cells))
         csrBoundaryChain = csrBoundaryMat * csrChain
         for k,value in enumerate(csrBoundaryChain.data):
              if value % 2 == 0: csrBoundaryChain.data[k] = 0
         out = [k for k,val in enumerate(csrBoundaryChain.data.tolist()) if val == 1]
         return out
Macro referenced in 49b.
\langle Test examples of From cells and facets to boundary cells 30b\rangle \equiv
     boundaryCells_2 = boundaryCells(CV,FV)
     boundaryCells_1 = boundaryCells([FV[k] for k in boundaryCells_2],EV)
     print "\nboundaryCells_2 =\n", boundaryCells_2
     print "\nboundaryCells_1 =\n", boundaryCells_1
     boundaryModel = (V,[FV[k] for k in boundaryCells_2])
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(boundaryModel)))
Macro referenced in 50.
```

3.3.2 Oriented operators

Two d-cells are said coherently oriented when their common (d-1)-facet has opposite orientations with respect to the two cells. When the boundary of an orientable solid partitionates its affine hull in two subsets corresponding to the *interior* and the *exterior* of the solid, then the boundary cells can be coherently oriented. This task is performed by the function signedBoundaryCells and signedCellularBoundaryCells in the following scripts. The sparse matricial structures returned by the functions signedSimplicialBoundary and signedCellularBoundary take values in the Abelian group $\{-1,0,1\}$. We call them signed matrices, and call signed operators the corresponding boundary and coboundary.

Signed boundary matrix for simplicial complexes The computation of the signed boundary matrix for simplicial complexes starts with enumerating the non-zero elements of the mod two (unoriented) boundary matrix. In particular, the pairs variable contains all the pairs of incident ((d-1)-cell, d-cell), corresponding to each 1 elements in the binary boundary matrix. Of course, their number equates the product of the number of d-cells, times the number of (d-1)-facets on the boundary of each d-cell.

For the case of a 3-simplicial complex CV, we have 4|CV| pairs elements. The actual goal of the function signedSimplicialBoundary, in the macro below, is to compute a sign for each of them.

The pairs values must be interpreted as (i, j) values in the incidence matrix FC (facets-cells), and hence as pairs of indices f and c into the characteristic matrices FV = CSR(M_{d-1}) and CV = CSR(M_d), respectively.

For each incidence pair f,c, the list vertLists contains the two lists of vertices associated to f and to c, called respectively the face and the coface. For each face, coface pair (i.e. for each unit element in the unordered boundary matrix), the missingVertIndices list will contain the index of the coface vertex not contained in the incident face.

Finally, the ± 1 (signed) incidence coefficients are computed and stored in the faceSigns, and then located in their actual positions within the csrSignedBoundaryMat. The sign of the incidence coefficient associated to the pair (facet,cell), also called (face,coface) in the implementation below, is computed as the sign of $(-1)^k$, where k is the position index of the removed vertex in the facet $\langle v_0, \ldots, v_{k-1}, v_{k+1}, \ldots, v_d \rangle$. of the $\langle v_0, \ldots, v_d \rangle$ cell.

```
\langle Signed boundary matrix for simplicial models 31 \rangle \equiv
     def signedSimplicialBoundary (CV,FV):
         # compute the set of pairs of indices to [boundary face,incident coface]
         coo = boundary(CV,FV).tocoo()
         pairs = [[coo.row[k],coo.col[k]] for k,val in enumerate(coo.data) if val != 0]
         # compute the [face, coface] pair as vertex lists
         vertLists = [[FV[f], CV[c]] for f,c in pairs]
         # compute the local (interior to the coface) indices of missing vertices
         def missingVert(face,coface): return list(set(coface).difference(face))[0]
         missingVertIndices = [c.index(missingVert(f,c)) for f,c in vertLists]
         # signed incidence coefficients
         def checkPermutation(vertLists,missingVertIndices):
             sameOrientation = []
             for (face,coface),index in zip(vertLists,missingVertIndices):
                 cell = tuple([vert for k,vert in enumerate(coface) if k!=index])
                 if len(cell)==2 and cell==face: sameOrientation += [1]
                 else: sameOrientation += [-1] # TODO: generalize for any "cell"
             return sameOrientation
```

```
sameOrientation = checkPermutation(vertLists,missingVertIndices)
faceSigns = AA(C(POWER)(-1))(missingVertIndices)
faceSigns = AA(PROD)(TRANS([faceSigns,sameOrientation]))

# signed boundary matrix
csrSignedBoundaryMat = csr_matrix( (faceSigns, TRANS(pairs)) )
return csrSignedBoundaryMat
```

Macro referenced in 49b.

Computation of signed boundary simplices The matrix of the signed boundary operator, with elements in $\{-1,0,1\}$, is computed in compressed sparse row (CSR) format, and stored in csrSignedBoundaryMat. In order to be able to return a list of signedBoundaryCells having a coherent orientation, we need to compute the coface of each boundary facet, i.e. the single d-cell having the facet on its boundary, and provide a coherent orientation to such chain of d-cells. The goal is obtained computing the sign of the determinant of the coface matrices, i.e. of square matrices having as rows the vertices of a coface, in normalised homogeneous coordinates.

The chain of boundary facets boundaryCells, obtained by multiplying the signed matrix of the boundary operator by the coordinate representation of the total *d*-chain, is coherently oriented by multiplication times the determinants of the cofaceMats.

The cofaceMats list is filled with the matrices having per row the position vectors of vertices of a coface, in normalized homogeneous coordinates. The list of signed face indices orientedBoundaryCells is returned by the function.

```
\langle Orientation of general convex cells 32 \rangle \equiv
     def swap(mylist): return [mylist[1]]+[mylist[0]]+mylist[2:]
     def boundaryCellsCocells(cells,facets):
         csrSignedBoundaryMat = signedSimplicialBoundary(V,cells,facets)
         csrTotalChain = totalChain(cells)
         csrBoundaryChain = matrixProduct(csrSignedBoundaryMat, csrTotalChain)
         cooCells = csrBoundaryChain.tocoo()
         boundaryCells = []
         for k,v in enumerate(cooCells.data):
             if abs(v) == 1:
                 boundaryCells += [int(cooCells.row[k] * cooCells.data[k])]
         boundaryCocells = []
         for k,v in enumerate(boundaryCells):
             boundaryCocells += list(csrSignedBoundaryMat[abs(v)].tocoo().col)
         return boundaryCells,boundaryCocells
     def signedBoundaryCells(verts,cells,facets):
         boundaryCells,boundaryCocells = boundaryCellsCocells(cells,facets)
```

```
boundaryCofaceMats = [[verts[v]+[1] for v in cells[c]] for c in boundaryCocells]
boundaryCofaceSigns = AA(SIGN)(AA(np.linalg.det)(boundaryCofaceMats))
orientedBoundaryCells = list(array(boundaryCells)*array(boundaryCofaceSigns))
return orientedBoundaryCells

Amacro referenced in 49b.
```

Signed boundary matrix for polytopal complexes

```
\langle Signed boundary matrix for polytopal complexes 33a\rangle \equiv
     """ Signed boundary matrix for polytopal complexes """
     def signedCellularBoundary(V,bases):
         coo = boundary(bases[-1],bases[-2]).tocoo()
         pairs = [[coo.row[k],coo.col[k]] for k,val in enumerate(coo.data) if val != 0]
         signs = []
         dim = len(bases)-1
         chain = incidenceChain(bases)
         for pair in pairs:
                                    # for each facet/coface pair
             flag = REVERSE(pair) # [c,f]
             #print "flag 1 =",flag
             for k in range(dim-1):
                 cell = flag[-1]
                 flag += [chain[k+1][cell][1]]
             verts = [CCOMB([V[v] for v in bases[dim-k][flag[k]]]) for k in range(dim+1)]
             flagMat = [verts[v]+[1] for v in range(dim+1)]
             flagSign = SIGN(np.linalg.det(flagMat))
             signs += [flagSign]
         csrSignedBoundaryMat = csr_matrix( (signs, TRANS(pairs)) )
         # numpy.set_printoptions(threshold=numpy.nan)
         # print csrSignedBoundaryMat.todense()
         return csrSignedBoundaryMat
```

Macro referenced in 49b.

Oriented boundary cells for polytopal complexes

```
⟨ Signed boundary cells for polytopal complexes 33b⟩ ≡
   """ Signed boundary cells for polytopal complexes """
   from scipy.sparse import *
   def signedCellularBoundaryCells(verts,bases):
```

```
CV = bases[-1]
boundaryMat = signedCellularBoundary(verts,bases)
chainCoords = csc_matrix((len(CV), 1))
for cell in range(len(CV)): chainCoords[cell,0] = 1
boundaryCells = list((boundaryMat * chainCoords).tocoo().row)
orientations = list((boundaryMat * chainCoords).tocoo().data)
return orientations,boundaryCells
```

Macro referenced in 49b.

3.3.3 Examples

Boundary of a 2D cuboidal grid The larCuboids function, when applied to a shape parameter and to the optional parameter full=True, returns both the intoger vertices V of the generated complex, and the list of bases of cells of dimension k ($0 \le k \le d$), where d = len(shape) - 1.

Oriented cuboidal and simplicial cells In the example test/py/larcc/test15.py we generate a simplicial and a cuboidal decomposition of the space parallelepiped with shape = [5,5,3]. In both cases the boundary matrix is computed by using the general polytopal approach provided by the signedCellularBoundaryCells function, showing in both cases the oriented boundary of the two complexes (Just notice that in the cuboidal version pyplasm makes a wrong rendering, to be fixed).

```
"test/py/larcc/test15.py" 34b \equiv """ Oriented cuboidal and simplicial cells (same algorithm) """ from larlib import *
```

```
# cuboidal grid
     V,bases = larCuboids([5,5,3],True)
     [VV,EV,FV,CV] = bases
     orientedBoundary = signedCellularBoundaryCells(V,AA(AA(REVERSE))([VV,EV,FV,CV]))
     cells = [FV[f] if sign==1 else REVERSE(FV[f]) for (sign,f) in zip(*orientedBoundary)]
     VIEW(EXPLODE(1.25,1.25,1.25)(MKPOLS((V,cells))))
     # simplicial grid
     V,CV = larSimplexGrid1([5,5,3])
     FV = larSimplexFacets(CV)
     EV = larSimplexFacets(FV)
     VV = AA(LIST)(range(len(V)))
     bases = [VV,EV,FV,CV]
     orientedBoundary = signedCellularBoundaryCells(V,bases)
     cells = [FV[f] if sign==1 else REVERSE(FV[f]) for (sign,f) in zip(*orientedBoundary)]
     VIEW(EXPLODE(1.25,1.25,1.25)(MKPOLS((V,cells))))
"test/py/larcc/test18.py" 35 \equiv
     """ Oriented cuboidal cells """
     """ Oriented cuboidal cells """
     from larlib import *
     def orientedBoundaryCells(V,(VV,EV,FV,CV)):
         boundaryMat = signedCellularBoundary(V,[VV,EV,FV,CV])
         chainCoords = csc_matrix((len(CV), 1))
         for cell in range(len(CV)): chainCoords[cell,0] = 1
         boundaryCells = list((boundaryMat * chainCoords).tocoo().row)
         orientations = list((boundaryMat * chainCoords).tocoo().data)
         return zip(orientations,boundaryCells)
     def normalVector(V,facet):
         v0, v1, v2 = facet[:3]
         return VECTPROD([ DIFF([V[v1],V[v0]]), DIFF([V[v2],V[v0]]) ])
     # cuboidal grid
     V,bases = larCuboids([5,5,3],True)
     [VV,EV,FV,CV] = bases
     BCpairs = orientedBoundaryCells(V,[VV,EV,FV,CV])
     orientedBoundary = [FV[face] if sign>0 else swap(FV[face]) for (sign,face) in BCpairs]
     normals = [ normalVector(V,facet) for facet in orientedBoundary ]
     facetCentroids = [CCOMB([V[v] for v in facet]) for facet in orientedBoundary]
     appliedNormals = [[centroid,SUM([centroid,normal])] for (centroid,normal) in zip(facetCentroid
     normalVectors = AA(POLYLINE)(appliedNormals)
     orientedQuads = [[sign,FV[face]] if sign>0 else [sign,swap(FV[face])] for (sign,face) in BCpai
```

3.3.4 Boundary orientation of a random (2D) cubical complex

```
"test/py/larcc/test17.py" 36 \equiv
     """ Boundary orientation of a random 2D cubical complex """
     from larlib import *
     from random import random
     # test model generation
     shape = 20,20
     V,FV = larCuboids(shape)
     cellSpan = prod(shape)
     fraction = 0.5
     remove = [int(random()*cellSpan) for k in range(int(cellSpan*fraction)) ]
     FV = [FV[k] for k in range(cellSpan) if not (k in remove)]
     _,EV = larCuboidsFacets((V,FV))
     VV = AA(LIST)(range(len(V)))
     orientedBoundary = signedCellularBoundaryCells(V,[VV,EV,FV])
     cells = [EV[e] if sign==1 else REVERSE(EV[e]) for (sign,e) in zip(*orientedBoundary)]
     # test model visualization
     VIEW(STRUCT(MKPOLS((V,FV))))
     VIEW(STRUCT(MKPOLS((V,EV))))
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,cells))))
     VIEW(STRUCT(MKPOLS((V,cells))))
     VIEW(mkSignedEdges((V,cells),2))
```

3.3.5 Boundary orientation of a random (2D) triangulation

Here we provide a 2D example of computation of the oriented boundary of a quite convoluted random cellular complex. The steps performed by the scripts in the following paragraphs are listed below:

- 1. vertices are generated as random point in the unit circle
- 2. the Delaunay triangulation of the whole set of points is built.
- 3. spike-like triangles elimination
- 4. the 90% of triangles is randomly discarded

- 5. the input LAR is provided by the remaining triangles
- 6. the 1-cells are computed, and if $n_i < n_j$ oriented as $v_i \rightarrow v_j$
- 7. the 2-cells are "coherently oriented" via the sign of their 3x3 determinant using normalised homogeneous coordinates of vertices: ccw if det > 0
- 8. the signed boundary matrix $[\partial_2]$ is built (with elements in $\{-1,0,1\}$)
- 9. the signed boundary 1-chain (the red one) is computed by $[\partial_2][\mathbf{1}_2]$, where $[\mathbf{1}_2]$ is the coordinate representation of the total 2-chain

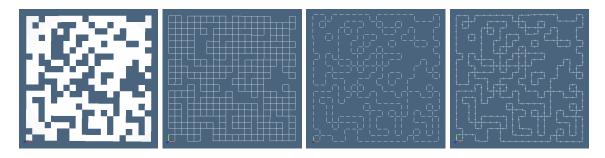


Figure 9: The orientation of the boundary of a random cuboidal 2-complex; (a) 2-cells; (b) 1-cells; (c) exploded boundary 1-chain; (d) oriented boundary 1-chain.

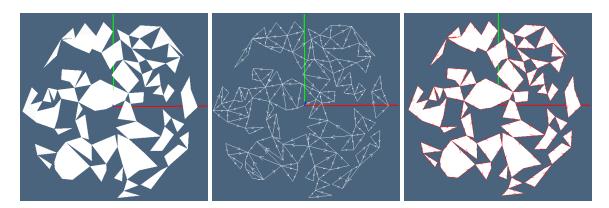


Figure 10: The orientation of the boundary of a random simplicial 2-complex; (a) 2-cells; (b) 1-cells; (c) oriented boundary 1-chain (red).

Top-down implementation

[&]quot;test/py/larcc/test16.py" $37 \equiv$

```
""" Boundary orientation of a random 2D triangulation """
from larlib import *
from random import random
from scipy import spatial,linalg

⟨ Vertices V generated as random point in the unit circle 38a⟩
⟨ Delaunay triangulation of the whole set V of points 38b⟩
⟨ Fraction of triangles randomly discarded 38c⟩
⟨ Coherently orient the input LAR model (V,FV) 39a⟩
⟨ Compute the 1-cell and 0-cell bases EV and VV 39b⟩
⟨ Signed 2-boundary matrix and signed boundary 1-chain 39c⟩
⟨ Display the boundary 1-chain 40a⟩

⋄
```

Vertices V generated as random point in the unit circle

Delaunay triangulation of the whole set V of points

```
⟨ Delaunay triangulation of the whole set V of points 38b⟩ ≡
    """ Delaunay triangulation of the whole set V of points """
    triangles = spatial.Delaunay(verts)
    def area(cell): return linalg.det([verts[v]+[1] for v in cell])/2
    cells = [ cell for cell in triangles.vertices.tolist() if area(cell)>PI/(3*npoints)]
    V, FV = AA(list)(verts), cells
    ⋄
```

Fraction of triangles randomly discarded

Macro referenced in 37.

Macro referenced in 37.

 \langle Fraction of triangles randomly discarded 38c $\rangle \equiv$

```
""" Fraction of triangles randomly discarded """
     fraction = 0.25
     cellSpan = len(FV)
     remove = [int(random()*cellSpan) for k in range(int(cellSpan*fraction)) ]
     FV = [FV[k] for k in range(cellSpan) if not k in remove]
Macro referenced in 37.
Coherent orientation of input LAR model (V,FV)
\langle Coherently orient the input LAR model (V,FV) 39a\rangle \equiv
     """ Coherently orient the input LAR model (V,FV) """
     def positiveOrientation(model):
         V, simplices = model
         out = []
         for simplex in simplices:
              theMat = [V[v]+[1] for v in simplex]
              if sign(linalg.det(theMat)) > 0: out += [simplex]
              else: out += [REVERSE(simplex)]
         return V, out
     V,FV = positiveOrientation((V,FV))
Macro referenced in 37.
Compute the 1-cell and 0-cell bases EV and VV
\langle Compute the 1-cell and 0-cell bases EV and VV 39b \rangle \equiv
     """ Compute the 1-cell and 0-cell bases EV and VV """
     EV = larSimplexFacets(FV)
     VV = AA(LIST)(range(len(V)))
     VIEW(mkSignedEdges((V,EV)))
Macro referenced in 37.
Signed boundary matrix [\partial_2] and signed boundary 1-chain
\langle Signed 2-boundary matrix and signed boundary 1-chain 39c\rangle \equiv
     """ Signed 2-boundary matrix and signed boundary 1-chain """
     orientedBoundary = signedCellularBoundaryCells(V,[VV,EV,FV])
     cells = [EV[e] if sign==1 else REVERSE(EV[e]) for (sign,e) in zip(*orientedBoundary)]
```

Macro referenced in 37.

Display the boundary 1-chain

3.4 Orienting polytopal cells

An orientation can be allocated to a general convex (polytopal) cell by computing the biggest simplex in its interior, and attributing to the cell the orientation of the contained simplex. It is in fact easy to see that the orientation can be propagated via adjacent coherently oriented simplexes, until to cover the whole cell.

The variables in the following script have the meaning specified below: input: "cell" indices of a convex and solid polytopes and "V" vertices; output: biggest "simplex" indices spanning the polytope; m: number of cell vertices; d: dimension (number of coordinates) of cell vertices; d+1: number of simplex vertices; vcell: cell vertices; vsimplex: simplex vertices; d: identity matrix; d: orthonormal spanning set of vectors e_k ; d: position vector of a simplex vertex in translated coordinates; d: unUsedIndices: cell indices not moved to simplex.

```
\langle Oriented boundary cells for simplicial models 40b\rangle \equiv
     def pivotSimplices(V,CV,d=3):
         simplices = []
         for cell in CV:
             vcell = np.array([V[v] for v in cell])
             m, simplex = len(cell), []
              # translate the cell: for each k, vcell[k] -= vcell[0], and simplex[0] := cell[0]
             for k in range(m-1,-1,-1): vcell[k] -= vcell[0]
              \# simplex = [0], basis = [], tensor = Id(d+1)
              simplex += [cel1[0]]
              basis = []
              tensor = np.array(IDNT(d))
              # look for most distant cell vertex
             dists = [SUM([SQR(x) for x in v])**0.5 for v in vcell]
             maxDistIndex = max(enumerate(dists),key=lambda x: x[1])[0]
             vector = np.array([vcell[maxDistIndex]])
              # normalize vector
              den=(vector**2).sum(axis=-1) **0.5
             basis = [vector/den]
              simplex += [cell[maxDistIndex]]
```

```
unUsedIndices = [h for h in cell if h not in simplex]
        # for k in \{2,d+1\}:
        for k in range((2,d+1)):
            # update the orthonormal tensor
            e = basis[-1]
            tensor = tensor - np.dot(e.T, e)
            # compute the index h of a best vector
            # look for most distant cell vertex
            dists = [SUM([SQR(x) for x in np.dot(tensor,v)])**0.5
            if h in unUsedIndices else 0.0
            for (h,v) in zip(cell,vcell)]
            # insert the best vector index h in output simplex
            maxDistIndex = max(enumerate(dists), key=lambda x: x[1])[0]
            vector = np.array([vcell[maxDistIndex]])
            # normalize vector
            den=(vector**2).sum(axis=-1) **0.5
            basis += [vector/den]
            simplex += [cell[maxDistIndex]]
            unUsedIndices = [h for h in cell if h not in simplex]
        simplices += [simplex]
    return simplices
def simplexOrientations(V,simplices):
    vcells = [[V[v]+[1.0] \text{ for } v \text{ in simplex}] \text{ for simplex in simplices}]
    return [SIGN(np.linalg.det(vcell)) for vcell in vcells]
```

Macro referenced in 49b.

Affine transformations of points Some primitive maps of points to points are given in the following, including translation, rotation and scaling of array of points via direct transformation of their coordinates. Second-order functions are used in order to employ their curried version to transform geometric assemblies.

```
⟨ Affine transformations of d-points 41⟩ ≡

def larTranslate (tvect):
    def larTranslate0 (points):
        return [VECTSUM([p,tvect]) for p in points]
    return larTranslate0

def larRotate (angle):  # 2-dimensional !! TODO: n-dim
    def larRotate0 (points):
        a = angle
        return [[x*COS(a)-y*SIN(a), x*SIN(a)+y*COS(a)] for x,y in points]
    return larRotate0
```

```
def larScale (svect):
    def larScale0 (points):
        print "\n points =",points
        print "\n svect =",svect
        return [AA(PROD)(TRANS([p,svect])) for p in points]
    return larScale0
```

Macro referenced in 49b.

4 Piecewise-linear mapping of topological spaces

A very simple but foundational software subsystem is developed in this section, by giving a general mechanism to produce curved maps of topological spaces, via the simplicial decomposition of a chart, i.e. of a planar embedding of the fundamental polygon of a d-dimensional manifold, and the definition of coordinate functions to be applied to its vertices (0-cells of the decomposition) to generate an embedding of the manifold.

4.1 Domain decomposition

A simplicial map is a map between simplicial complexes with the property that the images of the vertices of a simplex always span a simplex. Simplicial maps are thus determined by their effects on vertices, and provide a piecewise-linear approximation of their underlying polyhedra.

Since double simmeries are always present in the curved primitives generated in the module, an alternative cellular decomposition with cuboidal cells is provided. The default choice is "cuboid".

Standard and scaled decomposition of unit domain The larDomain of given shape is decomposed by larSimplexGrid1 as an hypercube of dimension $d \equiv len(shape)$, where the shape tuple provides the number or row, columns, pages, etc. of the decomposition.

```
⟨Generate a simplicial decomposition of the [0,1]^d domain 42⟩ ≡

""" cellular decomposition of the unit d-cube """

def larDomain(shape, cell='cuboid'):

   if cell=='simplex': V,CV = larSimplexGrid1(shape)
   elif cell=='cuboid': V,CV = larCuboids(shape)

   V = larScale( [1./d for d in shape])(V)
   return [V,CV]

\diamond
```

Macro referenced in 49b.

A scaled simplicial decomposition is provided by the second-order larIntervals function, with len(shape) and len(size) parameters, where the d-dimensionale vector len(size) is assumed as the scaling vector to be applied to the point $\mathbf{1} \in \mathbb{E}^d$.

```
⟨Scaled simplicial decomposition of the [0,1]<sup>d</sup> domain 43a⟩ ≡

def larIntervals(shape, cell='cuboid'):

def larIntervals0(size):

V,CV = larDomain(shape,cell)

V = larScale( size)(V)

return [V,CV]

return larIntervals0

⋄

Macro referenced in 49b.
```

4.2 Mapping domain vertices

The second-order textttlarMap function is the LAR implementation of the PLaSM primitive MAP. It is applied to the array coordFuncs of coordinate functions and to the simplicially decomposed domain, returning an embedded and/or curved domain instance.

4.3 Identify close or coincident points

Macro referenced in 49b.

The function checkModel, applied to a model parameter, i.e. to a (vertices, cells) pair, returns the model after identification of vertices with coincident or very close position vectors. The checkModel function works as follows: first a dictionary vertDict is created, with key a suitably approximated position converted into a string by the vcode converter (given in the Appendix), and with value the list of vertex indices with the same (approximated) position. Then, an invertedindex array is created, associating each original vertex index with the new index produced by enumerating the (distinct) keys of the dictionary. Finally, a new list CV of cells is created, by substituting the new vertex indices for the old ones.

 \langle Create a dictionary with key the point location $43c \rangle \equiv$

```
from collections import defaultdict
     def checkModel(model,dim=2):
         V,CV = model; n = len(V)
         vertDict = defaultdict(list)
         for k,v in enumerate(V): vertDict[vcode(v)].append(k)
         points,verts = TRANS(vertDict.items())
         invertedindex = [None]*n
         V = []
         for k,value in enumerate(verts):
             V.append(eval(points[k]))
             for i in value:
                 invertedindex[i]=k
         CV = [[invertedindex[v] for v in cell] for cell in CV]
         # filter out degenerate cells
         CV = [list(set(cell)) for cell in CV if len(set(cell))>=dim+1]
         return [V, CV]
Macro referenced in 49b.
```

Macro referenced in 44a, 46a.

2-cell with high topological genus

```
"test/py/larcc/test23.py" 44a \equiv
      """ 2-cell with high topological genus """
      from larlib import *
      ⟨Output of 2-cell with high topological genus 0. 44b⟩
      (Output of 2-cell with high topological genus 1. 45a)
      (Output of 2-cell with high topological genus 2. 45b)
      \Diamond
\langle \text{Output of 2-cell with high topological genus 0. 44b} \rangle \equiv
      """ 2-cell with high topological genus 0. """
      side = QUOTE(5*[1,-1])
      holes = PROD([side, side])
      V,FV,_ = UKPOL(holes)
      FV = [[v-1 \text{ for } v \text{ in } f] \text{ for } f \text{ in } FV]
      EV = CAT([[[v,f[(k+1)\%4]] \text{ for } k,v \text{ in enumerate}(f+[f[0]][:-1])] \text{ for } f \text{ in } FV])
      VIEW(STRUCT(MKPOLS((V,EV))))
      (W,_) = larBox((-1,-1),(10,10))
      (_,(_,EW,FW)) = larCuboids((1,1),True)
      complex = Struct([(W,FW,EW),(V,FV,EV)])
      V,FV,EV = struct2lar(complex)
```

```
⟨Output of 2-cell with high topological genus 1. 45a⟩ ≡
    """ 2-cell with high topological genus 1. """
FV = [sorted(CAT(FV))]
# EV = sorted(AA(sorted)(EV)) # p2t bug if uncommented: CHECK
VIEW(STRUCT(MKPOLS((V,EV))))

VV = AA(LIST)(range(len(V)))
submodel = STRUCT(MKPOLS((V,EV)))
VIEW(larModelNumbering(1,1,1)(V,[VV,EV,FV],submodel,1.5))
⋄
```

Macro referenced in 44a.

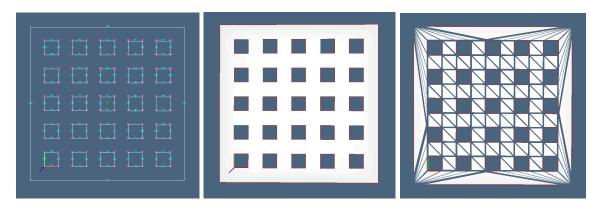


Figure 11: LAR of a 2-complex: (a) numbering of complex 0-, 1- and 2-cells (just one, numbered 0); (b) solid display of the complex, with boundary 1-cells (all 1-cells) colored in red; (c) triangulation of the interior, made for display purpose.

Explicit LAR of 2-cell with high topological genus The explicit definition of the LAR of the 2-complex shown in Figura 12 is given below. The reader should notice that the LAR contains the 2-faces description EV of a single 2-cell—of cardinality O(V), and that the bulk of this representation is the array of edge indices EV, given here in lexicographic order.

```
"test/py/larcc/test24.py" 46a \equiv
     """ Explicit LAR of 2-cell with high topological genus """
     from larlib import *
     (Output of 2-cell with high topological genus 0. 44b)
     FV = [CAT(FV)] + sorted(AA(list)(FV))[1:]
     # EV = sorted(AA(sorted)(EV)) # p2t bug if uncommented: CHECK
     model = V,FV,EV
     VV = AA(LIST)(range(len(V)))
     submodel = STRUCT(MKPOLS((V,EV)))
     VIEW(larModelNumbering(1,1,1)(V,[VV,EV,FV],submodel,1.5))
     # LAR data structures
     chain = \lceil 1 \rceil + \lceil 0 \rceil * (len(FV) - 1)
     outModel,triangleSet = larComplexChain(model)(chain)
     viewLarComplexChain(model)([0]+[1]*(len(FV)-1))
     viewLarComplexChain(model)([1]+[0]*(len(FV)-1))
     viewLarComplexChain(model)([1]+[0]*5 + [1]*(len(FV)-6))
     viewLarComplexChain(model)([1]*(len(FV)))
"test/py/larcc/test25.py" 46b \equiv
     """ 2-cell with high topological genus """
     from larlib import *
     V = [[-1.0, -1.0], [-1.0, 10.0], [10.0, -1.0], [10.0, 10.0], [0.0, 1.0], [0.0, 0.0], [1.0, 0.0],
     [1.0,1.0],[0.0,3.0],[0.0,2.0],[1.0,2.0],[1.0,3.0],[0.0,5.0],[0.0,4.0],[1.0,4.0],
     [1.0,5.0],[0.0,7.0],[0.0,6.0],[1.0,6.0],[1.0,7.0],[0.0,9.0],[0.0,8.0],[1.0,8.0],
     [1.0,9.0], [2.0,1.0], [2.0,0.0], [3.0,0.0], [3.0,1.0], [2.0,3.0], [2.0,2.0], [3.0,2.0],
     [3.0,3.0], [2.0,5.0], [2.0,4.0], [3.0,4.0], [3.0,5.0], [2.0,7.0], [2.0,6.0], [3.0,6.0],
     [3.0,7.0],[2.0,9.0],[2.0,8.0],[3.0,8.0],[3.0,9.0],[4.0,1.0],[4.0,0.0],[5.0,0.0],
     [5.0,1.0], [4.0,3.0], [4.0,2.0], [5.0,2.0], [5.0,3.0], [4.0,5.0], [4.0,4.0], [5.0,4.0],
     [5.0,5.0], [4.0,7.0], [4.0,6.0], [5.0,6.0], [5.0,7.0], [4.0,9.0], [4.0,8.0], [5.0,8.0],
     [5.0,9.0], [6.0,1.0], [6.0,0.0], [7.0,0.0], [7.0,1.0], [6.0,3.0], [6.0,2.0], [7.0,2.0],
     [7.0,3.0],[6.0,5.0],[6.0,4.0],[7.0,4.0],[7.0,5.0],[6.0,7.0],[6.0,6.0],[7.0,6.0],
     [7.0,7.0],[6.0,9.0],[6.0,8.0],[7.0,8.0],[7.0,9.0],[8.0,1.0],[8.0,0.0],[9.0,0.0],
     [9.0,1.0],[8.0,3.0],[8.0,2.0],[9.0,2.0],[9.0,3.0],[8.0,5.0],[8.0,4.0],[9.0,4.0],
```



Figure 12: LAR of a 2-complex. Upper row: interior (white) and boundary (red) of different 2-chains; Lower row: triangulations of the chains. Warning: all extracted from LAR of a complex with the same boundary operator ... :o)

```
[9.0,5.0],[8.0,7.0],[8.0,6.0],[9.0,6.0],[9.0,7.0],[8.0,9.0],[8.0,8.0],[9.0,8.0],
[9.0,9.0]]
```

 $FV = \begin{bmatrix} [0,1,2,3,4,5,6,7,8,9,10,11,12,13,14,15,16,17,18,19,20,21,22,23,24,25,26,27,28,29,30,31,32,33,34,35,36,37,38,39,40,41,42,43,44,45,46,47,48,49,50,51,52,53,54,55,56,57,58,59,60,61,62,63,64,65,66,67,68,69,70,71,72,73,74,75,76,77,78,79,80,81,82,83,84,85,86,87,88,89,90,91,92,93,94,95,96,97,98,99,100,101,102,103], [4,5,6,7], [8,9,10,11], [12,13,14,15], [16,17,18,19], [20,21,22,23], [24,25,26,27], [28,29,30,31], [32,33,34,35], [36,37,38,39], [40,41,42,43], [44,45,46,47], [48,49,50,51], [52,53,54,55], [56,57,58,59], [60,61,62,63], [64,65,66,67], [68,69,70,71], [72,73,74,75], [76,77,78,79], [80,81,82,83], [84,85,86,87], [88,89,90,91], [92,93,94,95], [96,97,98,99], [100,101,102,103]]$

$$\begin{split} \text{EV} &= \left[(54,55), (92,93), (68,71), (94,95), (40,41), (18,19), (74,75), (16,19), (80,81), (89,90), (6,7), (52,53), (24,27), (25,26), (36,37), (33,34), (36,39), (84,85), (29,30), (52,55), (14,15), (97,98), (96,97), (32,35), (17,18), (72,73), (102,103), (101,102), (45,46), (46,47), (22,23), (69,70), (8,9), (64,67), (65,66), (58,59), (12,13), (68,69), (32,33), (37,38), (73,74), (50,51), (100,101), (38,39), (77,78), (56,59), (44,47), (49,50), (24,25), (76,77), (81,82), (66,67), (0,2), (34,35), (60,63), (1,3), (42,43), (28,29), (82,83), (13,14), (48,51), (61,62), (16,17), (93,94), (88,91), (48,49), (92,95), (40,43), (12,15), (41,42), (2,3), (90,91), (80,83), (84,87), (76,79), (53,54), (96,99), (85,86), (30,31), (62,63), (4,7), (10,11), (5,6), (28,31), (98,99), (57,58), (70,71), (56,57), (4,5), (0,1), (72,75), (60,61), (8,11), (9,10), (78,79), (44,45), (86,87), (64,65), (20,23), (26,27), (21,22), (100,103), (88,89), (20,21) \end{split}$$

5 Exporting the library

5.1 MIT licence

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 \Diamond

Macro referenced in 49b.

5.2 Importing of modules or packages

```
⟨Importing of modules or packages 49a⟩ ≡
    from larlib import *

import collections
import numpy as np
from scipy import zeros,arange,mat,amin,amax,array
from scipy.sparse import vstack,hstack,csr_matrix,coo_matrix,lil_matrix,triu
```

Macro referenced in 49b.

5.3 Writing the library file

```
"larlib/larlib/larcc.py" 49b =

# -*- coding: utf-8 -*-

""" Basic LARCC library """

from larlib import *

\( \text{The MIT Licence 48} \)

\( \text{Importing of modules or packages 49a} \)

\( \text{Affine transformations of $d$-points 41} \)
```

```
(From list of triples to scipy.sparse 3b)
       Brc to Coo transformation 2
       Coo to Csr transformation 3c >
       Brc to Csr transformation 4a
       Query Matrix shape 5a
       Sparse to dense matrix transformation 6a
      (Matrix product and transposition 6c)
      (Matrix filtering to produce the boundary matrix 8)
      (Matrix filtering via a generic predicate 9b)
       Characteristic matrix transposition 17
       From cells and facets to boundary operator 29b
       From cells and facets to boundary cells 30a
       Computation of topological relation 28
       Signed boundary matrix for simplicial models 31
       Orientation of general convex cells 32 >
       Computation of cell adjacencies 18b
       Extraction of facets of a cell complex 18a
       Some incidence operators 25a
       Visualization of cell indices 20a, ... >
       Numbered visualization of a LAR model ?
       Drawing of oriented edges 21a
       Incidence chain computation 27a
       Signed boundary matrix for polytopal complexes 33a
       Signed boundary cells for polytopal complexes 33b
       Oriented boundary cells for simplicial models 40b
       Generate a simplicial decomposition of the [0,1]^d domain 42
       Scaled simplicial decomposition of the [0,1]^d domain 43a
       Create a dictionary with key the point location 43c
       Primitive mapping function 43b
      (From 2D chains to boundary chains 24)
      if __name__ == "__main__":
          \langle \text{ Test examples 50} \rangle
     Unit tests
\langle \text{ Test examples 50} \rangle \equiv
      (Test example of Brc to Coo transformation 3a)
      (Test example of Coo to Csr transformation 3d)
      ⟨ Test example of Brc to Csr transformation 4b⟩
      (Test examples of Query Matrix shape 5b)
      (Test examples of Sparse to dense matrix transformation 6b)
      (Test example of Matrix filtering to produce the boundary matrix 9a)
```

6

```
\label{eq:continuous} $$ \langle {\rm Test} {\rm \ example \ of \ Matrix \ filtering \ via \ a \ generic \ predicate \ 9c} \rangle$$ $$ \langle {\rm Test} {\rm \ examples \ of \ From \ cells \ and \ facets \ to \ boundary \ operator \ 29a} \rangle$$ $$ \langle {\rm Test} {\rm \ examples \ of \ Computation \ of \ cell \ adjacencies \ 19b} \rangle$$ $$ \langle {\rm Test} {\rm \ examples \ of \ Extraction \ of \ facets \ of \ a \ cell \ complex \ 19a} \rangle$$$ $$ \rangle$$
```

Macro referenced in 49b.

Comparing oriented and unoriented boundary

```
"test/py/larcc/test09.py" 51a \equiv
     """ comparing oriented boundary and unoriented boundary extraction on a simple example """
     import sys; sys.path.insert(0, 'lib/py/')
     from largrid import *
     from larcc import *
     V,CV = larSimplexGrid1([1,1,1])
     FV = larSimplexFacets(CV)
     orientedBoundary = signedBoundaryCells(V,CV,FV)
     orientedBoundaryFV = [FV[-k] if k<0 else swap(FV[k]) for k in orientedBoundary]
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,orientedBoundaryFV))))
     BF = boundaryCells(CV,FV)
     boundaryCellsFV = [FV[k] for k in BF]
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,boundaryCellsFV))))
"test/py/larcc/test12.py" 51b \equiv
     """ comparing edge orientation and oriented boundary extraction """
     import sys; sys.path.insert(0, 'lib/py/')
     from largrid import *
     from larcc import *
     V,FV = larSimplexGrid1([5,5])
     EV = larSimplexFacets(FV)
     VIEW(mkSignedEdges((V,EV)))
     orientedBoundary = signedBoundaryCells(V,FV,EV)
     orientedBoundaryEV = [EV[-k] if k<0 else swap(EV[k]) for k in orientedBoundary]
     VIEW(mkSignedEdges((V,orientedBoundaryEV)))
```

A Appendix: Tutorials

A.1 Model generation, skeleton and boundary extraction

```
"test/py/larcc/test01.py" 52a \equiv
      """ Model generation, skeleton and boundary extraction """
      from larlib import *
      (input of 2D topology and geometry data 52b)
      ⟨ characteristic matrices 52c ⟩
      (incidence matrix 52d)
      (boundary and coboundary operators 52e)
      (product of cell complexes 53a)
      (2-skeleton extraction 53b)
      (1-skeleton extraction 53c)
      (0-coboundary computation 54a)
      (1-coboundary computation 54b)
      (2-coboundary computation 54c)
      ⟨ boundary chain visualisation 54d ⟩
\langle \text{ input of 2D topology and geometry data 52b} \rangle \equiv
      # input of geometry and topology
      V2 = [[4,10],[8,10],[14,10],[8,7],[14,7],[4,4],[8,4],[14,4]]
      EV = [[0,1],[1,2],[3,4],[5,6],[6,7],[0,5],[1,3],[2,4],[3,6],[4,7]]
      FV = [[0,1,3,5,6],[1,2,3,4],[3,4,6,7]]
Macro referenced in 52a.
\langle characteristic matrices 52c \rangle \equiv
      # characteristic matrices
      csrFV = csrCreate(FV)
      csrEV = csrCreate(EV)
      print "\nFV =\n", csr2DenseMatrix(csrFV)
      print "\nEV =\n", csr2DenseMatrix(csrEV)
Macro referenced in 52a.
\langle incidence matrix 52d \rangle \equiv
      # product
      csrEF = matrixProduct(csrEV, csrTranspose(csrFV))
      print "\nEF =\n", csr2DenseMatrix(csrEF)
Macro referenced in 52a.
\langle boundary and coboundary operators 52e\rangle \equiv
```

```
# boundary and coboundary operators
     facetLengths = [csrCell.getnnz() for csrCell in csrEV]
     boundary = csrBoundaryFilter(csrEF,facetLengths)
     coboundary_1 = csrTranspose(boundary)
     print "\ncoboundary_1 =\n", csr2DenseMatrix(coboundary_1)
Macro referenced in 52a.
\langle product of cell complexes 53a\rangle \equiv
     # product operator
     mod_2D = (V2, FV)
     V1,topol_0 = [[0.],[1.],[2.]], [[0],[1],[2]]
     topol_1 = [[0,1],[1,2]]
     mod_OD = (V1, topol_O)
     mod_1D = (V1, topol_1)
     V3,CV = larModelProduct([mod_2D,mod_1D])
     mod_3D = (V3,CV)
     VIEW(EXPLODE(1.2,1.2,1.2)(MKPOLS(mod_3D)))
     print "\nk_3 =", len(CV), "\n"
Macro referenced in 52a.
\langle 2-skeleton extraction 53b \rangle \equiv
     # 2-skeleton of the 3D product complex
     mod_2D_1 = (V2, EV)
     mod_3D_h2 = larModelProduct([mod_2D,mod_0D])
     mod_3D_v2 = larModelProduct([mod_2D_1,mod_1D])
     _{,FV_h} = mod_{3D_h2}
     \_,FV_v = mod_3D_v2
     FV3 = FV_h + FV_v
     SK2 = (V3, FV3)
     VIEW(EXPLODE(1.2,1.2,1.2)(MKPOLS(SK2)))
     print "\nk_2 =", len(FV3), "\n""
Macro referenced in 52a.
\langle 1-skeleton extraction 53c \rangle \equiv
     # 1-skeleton of the 3D product complex
     mod_2D_0 = (V2,AA(LIST)(range(len(V2))))
     mod_3D_h1 = larModelProduct([mod_2D_1,mod_0D])
     mod_3D_v1 = larModelProduct([mod_2D_0,mod_1D])
     _{,EV_h} = mod_{3D_h1}
     _{,EV_v} = mod_{3D_v1}
     EV3 = EV_h + EV_v
     SK1 = (V3, EV3)
```

```
VIEW(EXPLODE(1.2,1.2,1.2)(MKPOLS(SK1)))
     print "\nk_1 =", len(EV3), "\n"
Macro referenced in 52a.
\langle 0-coboundary computation 54a \rangle \equiv
     # boundary and coboundary operators
     np.set_printoptions(threshold=sys.maxint)
     csrFV3 = csrCreate(FV3)
     csrEV3 = csrCreate(EV3)
     csrVE3 = csrTranspose(csrEV3)
     facetLengths = [csrCell.getnnz() for csrCell in csrEV3]
     boundary = csrBoundaryFilter(csrVE3,facetLengths)
     coboundary_0 = csrTranspose(boundary)
     print "\ncoboundary_0 =\n", csr2DenseMatrix(coboundary_0)
Macro referenced in 52a.
\langle 1-coboundary computation 54b \rangle \equiv
     csrEF3 = matrixProduct(csrEV3, csrTranspose(csrFV3))
     facetLengths = [csrCell.getnnz() for csrCell in csrFV3]
     boundary = csrBoundaryFilter(csrEF3,facetLengths)
     coboundary_1 = csrTranspose(boundary)
     print "\ncoboundary_1.T =\n", csr2DenseMatrix(coboundary_1.T)
Macro referenced in 52a.
\langle\, \text{2-coboundary computation 54c}\,\rangle \equiv
     csrCV = csrCreate(CV)
     csrFC3 = matrixProduct(csrFV3, csrTranspose(csrCV))
     facetLengths = [csrCell.getnnz() for csrCell in csrCV]
     boundary = csrBoundaryFilter(csrFC3,facetLengths)
     coboundary_2 = csrTranspose(boundary)
     print "\ncoboundary_2 =\n", csr2DenseMatrix(coboundary_2)
Macro referenced in 52a.
\langle boundary chain visualisation 54d\rangle \equiv
     # boundary chain visualisation
     boundaryCells_2 = boundaryCells(CV,FV3)
     boundary = (V3,[FV3[k] for k in boundaryCells_2])
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(boundary)))
Macro referenced in 52a.
```

A.2 Boundary of 3D simplicial grid

```
"test/py/larcc/test02.py" 55a \equiv
     """ Boundary of 3D simplicial grid """
     from larlib import *
     ⟨boundary of 3D simplicial grid 55b⟩
\langle boundary of 3D simplicial grid 55b\rangle \equiv
     V,CV = larSimplexGrid1([10,10,3])
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,CV))))
     SK2 = (V,larSimplexFacets(CV))
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(SK2)))
     _{,FV} = SK2
     SK1 = (V,larSimplexFacets(FV))
     _{,EV} = SK1
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(SK1)))
     boundaryCells_2 = boundaryCells(CV,FV)
     boundary = (V,[FV[k] for k in boundaryCells_2])
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(boundary)))
     print "\nboundaryCells_2 =\n", boundaryCells_2
     boundaryCells_2 = signedBoundaryCells(V,CV,FV)
     boundaryFV = [FV[-k] if k<0 else swap(FV[k]) for k in boundaryCells_2]
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,boundaryFV))))
     print "\nboundaryCells_2 =\n", boundaryFV
```

Macro referenced in 55a.

A.3 Oriented boundary of a random simplicial complex

```
"test/py/larcc/test03.py" 55c ≡

""" Oriented boundary of a random simplicial complex """

from larlib import *

⟨Importing external modules 55d⟩
⟨Generating and viewing a random 3D simplicial complex 56a⟩
⟨Computing and viewing its non-oriented boundary 56b⟩
⟨Computing and viewing its oriented boundary 56c⟩
⟨Importing external modules 55d⟩ ≡
```

```
""" Importing external modules """
     from scipy.spatial import Delaunay
     import numpy as np
Macro referenced in 55c.
\langle Generating and viewing a random 3D simplicial complex 56a\rangle \equiv
     verts = np.random.rand(10000, 3) # 1000 points in 3-d
     verts = [AA(lambda x: 2*x)(VECTDIFF([vert,[0.5,0.5,0.5]])) for vert in verts]
     verts = [vert for vert in verts if VECTNORM(vert) < 1.0]</pre>
     tetra = Delaunay(verts)
     cells = [cell for cell in tetra.vertices.tolist()
               if ((verts[cell[0]][2]<0) and (verts[cell[1]][2]<0)
                        and (verts[cel1[2]][2]<0) and (verts[cel1[3]][2]<0) ) ]
     V, CV = verts, cells
     VIEW(MKPOL([V,AA(AA(lambda k:k+1))(CV),[]]))
Macro referenced in 55c.
\langle Computing and viewing its non-oriented boundary 56b\rangle \equiv
     FV = larSimplexFacets(CV)
     VIEW(MKPOL([V,AA(AA(lambda k:k+1))(FV),[]]))
     boundaryCells_2 = boundaryCells(CV,FV)
     print "\nboundaryCells_2 =\n", boundaryCells_2
     bndry = (V,[FV[k] for k in boundaryCells_2])
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(bndry)))
Macro referenced in 55c.
\langle Computing and viewing its oriented boundary 56c\rangle \equiv
     boundaryCells_2 = signedBoundaryCells(V,CV,FV)
     print "\nboundaryCells_2 =\n", boundaryCells_2
     boundaryFV = [FV[-k] if k<0 else swap(FV[k]) for k in boundaryCells_2]
     boundaryModel = (V,boundaryFV)
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(boundaryModel)))
Macro referenced in 55c.
      Oriented boundary of a simplicial grid
"test/py/larcc/test04.py" 56d \equiv
     (Generate and view a 3D simplicial grid 57a)
     (Computing and viewing the 2-skeleton of simplicial grid 57b)
     (Computing and viewing the oriented boundary of simplicial grid 57c)
```

```
\langle Generate and view a 3D simplicial grid 57a\rangle \equiv
     """ Generate and view a 3D simplicial grid """
     from larlib import *
     V,CV = larSimplexGrid1([4,4,4])
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,CV))))
Macro referenced in 56d.
\langle Computing and viewing the 2-skeleton of simplicial grid 57b\rangle \equiv
     FV = larSimplexFacets(CV)
     EV = larSimplexFacets(FV)
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,FV))))
Macro referenced in 56d.
\langle Computing and viewing the oriented boundary of simplicial grid 57c\rangle \equiv
     csrSignedBoundaryMat = signedSimplicialBoundary (V,CV,FV)
     boundaryCells_2 = signedBoundaryCells(V,CV,FV)
     boundaryFV = [FV[-k] if k<0 else swap(FV[k]) for k in boundaryCells_2]
     boundary = (V,boundaryFV)
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(boundary)))
Macro referenced in 56d.
A.5 Skeletons and oriented boundary of a simplicial complex
"test/py/larcc/test05.py" 57d \equiv
     """ Skeletons and oriented boundary of a simplicial complex """
     from larlib import *
     (Skeletons computation and visualisation?)
      (Oriented boundary matrix visualization 58a)
     (Computation of oriented boundary cells 58b)
\langle Skeletons computation and vilualisation 57e\rangle \equiv
     """ Skeletons computation and vilualisation """
     V,FV = larSimplexGrid1([3,3])
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,FV))))
     EV = larSimplexFacets(FV)
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,EV))))
     VV = larSimplexFacets(EV)
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS((V,VV))))
```

Macro never referenced.

```
\langle Oriented boundary matrix visualization 58a\rangle \equiv
     """ Oriented boundary matrix visualization """
     np.set_printoptions(threshold='nan')
     csrSignedBoundaryMat = signedSimplicialBoundary (V,FV,EV)
     Z = csr2DenseMatrix(csrSignedBoundaryMat)
     print "\ncsrSignedBoundaryMat =\n", Z
     import matplotlib.pyplot
     from pylab import *
     matshow(Z)
     show()
Macro referenced in 57d.
\langle Computation of oriented boundary cells 58b\rangle \equiv
           Computation of oriented boundary cells """
     boundaryCells_1 = signedBoundaryCells(V,FV,EV)
     print "\nboundaryCells_1 =\n", boundaryCells_1
     boundaryEV = [EV[-k] if k<0 else swap(EV[k]) for k in boundaryCells_1]
     bndry = (V,boundaryEV)
     VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLS(bndry)))
Macro referenced in 57d.
```

A.6 Boundary of random 2D simplicial complex

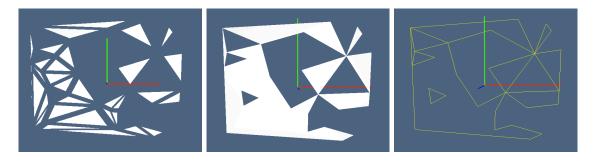


Figure 13: example caption

```
if m/a < 1.7 and m/b < 1.7 and m/c < 1.7: return True
          else: return False
Macro referenced in 58c.
\langle Generation and selection of random triangles 59a\rangle \equiv
     """ Generation and selection of random triangles """
     verts = np.random.rand(50,2)
     verts = (verts - [0.5, 0.5]) * 2
     triangles = Delaunay(verts)
     cells = [ cell for cell in triangles.vertices.tolist()
               if (not quasiEquilateral([verts[k] for k in cell])) ]
     V, FV = AA(list)(verts), cells
     EV = larSimplexFacets(FV)
     pols2D = MKPOLS((V,FV))
     VIEW(EXPLODE(1.5,1.5,1.5)(pols2D))
Macro referenced in 58c.
\langle Boundary computation and visualisation 59b\rangle \equiv
     """ Boundary computation and visualisation """
     orientedBoundary = signedBoundaryCells(V,FV,EV)
     submodel = mkSignedEdges((V,orientedBoundary))
     VIEW(submodel)
Macro referenced in 58c.
\langle Decompose a permutation into cycles 59c\rangle \equiv
     """ Decompose a permutation into cycles """
     def permutationOrbits(List):
          d = dict((i,int(x)) for i,x in enumerate(List))
          out = []
```

```
while d:
    x = list(d)[0]
    orbit = []
    while x in d:
        orbit += [x],
        x = d.pop(x)
    out += [CAT(orbit)+orbit[0]]
    return out

if __name__ == "__main__":
    print [2, 3, 4, 5, 6, 7, 0, 1]
    print permutationOrbits([2, 3, 4, 5, 6, 7, 0, 1])
    print [3,9,8,4,10,7,2,11,6,0,1,5]
    print permutationOrbits([3,9,8,4,10,7,2,11,6,0,1,5])
```

Macro never referenced.

A.7 Assemblies of simplices and hypercubes

```
"test/py/larcc/test07.py" 60a ≡

""" Assemblies of simplices and hypercubes """

from larlib import *

⟨Definition of 1-dimensional LAR models 60b⟩
⟨Assembly generation of squares and triangles 61a⟩
⟨Assembly generation of cubes and tetrahedra 61b⟩

⋄
```

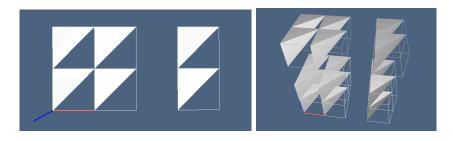


Figure 14: (a) Assemblies of squares and triangles; (b) assembly of cubes and tetrahedra.

```
⟨ Definition of 1-dimensional LAR models 60b⟩ ≡
    """ Definition of 1-dimensional LAR models """
    geom_0,topol_0 = [[0.],[1.],[2.],[3.],[4.]],[[0,1],[1,2],[3,4]]
    geom_1,topol_1 = [[0.],[1.],[2.]], [[0,1],[1,2]]
    mod_0 = (geom_0,topol_0)
    mod_1 = (geom_1,topol_1)
    ⋄
```

Macro referenced in 60a.

```
\langle Assembly generation of squares and triangles 61a\rangle \equiv
     """ Assembly generation of squares and triangles """
     squares = larModelProduct([mod_0,mod_1])
     V,FV = squares
     simplices = pivotSimplices(V,FV,d=2)
     VIEW(STRUCT([ MKPOL([V,AA(AA(C(SUM)(1)))(simplices),[]]),
                     SKEL_1(STRUCT(MKPOLS((V,FV)))) ]))
Macro referenced in 60a.
\langle Assembly generation of cubes and tetrahedra 61b \rangle \equiv
     """ Assembly generation of cubes and tetrahedra """
     cubes = larModelProduct([squares,mod_0])
     V,CV = cubes
     simplices = pivotSimplices(V,CV,d=3)
     VIEW(STRUCT([ MKPOL([V,AA(AA(C(SUM)(1)))(simplices),[]]),
                     SKEL_1(STRUCT(MKPOLS((V,CV)))) ]))
Macro referenced in 60a.
```

References

[CL13] CVD-Lab, *Linear algebraic representation*, Tech. Report 13-00, Roma Tre University, October 2013.