

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 $\text{BRC}(A)$ is used for the corresponding data structure.

$$A = \begin{pmatrix} 0, 1, 0, 0, 0, 0, 0, 1, 0, 0 \\ 0, 0, 1, 0, 0, 0, 0, 0, 0, 0 \\ 1, 0, 0, 1, 0, 0, 0, 0, 0, 1 \\ 1, 0, 0, 0, 0, 0, 1, 0, 0, 0 \\ 0, 0, 0, 0, 0, 1, 1, 1, 0, 0 \\ 0, 0, 1, 0, 1, 0, 0, 0, 1, 0 \\ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0 \\ 0, 1, 0, 0, 0, 0, 0, 1, 0, 1 \\ 0, 0, 0, 1, 0, 0, 0, 0, 1, 0 \\ 0, 1, 1, 0, 1, 0, 0, 0, 0, 0 \end{pmatrix} \mapsto \text{BRC}(A) = \begin{array}{l} [[1,7], \\ [2], \\ [0,3,9], \\ [0,6], \\ [5,6,7], \\ [2,4,8], \\ [], \\ [1,7,9], \\ [3,8], \\ [1,2,4]] \end{array}$$

1.2 Format conversions

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

```

⟨Brc to Coo transformation 2⟩ ≡
def brc2Coo(ListOfListOfInt):
    COOm = [[k,col,1] for k,row in enumerate(ListOfListOfInt)
              for col in row ]
    return COOm

```

Macro referenced in [33a](#).

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 [33b](#).

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 [33a](#).

The conversion from triples to `csr` format is provided below.

⟨ Coo to Csr transformation 3c ⟩ ≡

```
def coo2Csr(COOm):
    CSRm = triples2mat(COOm,"csr")
    return CSRm
```

◇

Macro referenced in [33a](#).

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

⟨ Test example of Coo to Csr transformation 3d ⟩ ≡

```

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

```

Macro referenced in 33b.

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

```

csr(FV) = <4x6 sparse matrix of type '<type 'numpy.int64'>'
  with 12 stored elements in Compressed Sparse Row format>
csr(EV) = <9x6 sparse matrix of type '<type 'numpy.int64'>'
  with 18 stored elements in Compressed Sparse Row format>

```

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 33a.

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

```

```
VIEW(STRUCT(MKPOLS((V,FV))))
VIEW(STRUCT(MKPOLS((V,EV))))
```

◇

Macro referenced in 7, 33b.



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*).

```
< Query Matrix shape 5a > ≡
def csrGetNumberOfRows(CSRmatrix):
    Int = CSRmatrix.shape[0]
    return Int

def csrGetNumberOfColumns(CSRmatrix):
    Int = CSRmatrix.shape[1]
    return Int
◇
```

Macro referenced in 33a.

```
< Test examples of Query Matrix shape 5b > ≡
```

```

print "\n>>> csrGetNumberOfRows"
print "\ncsrGetNumberOfRows(csrFV) =", csrGetNumberOfRows(csrFV)
print "\ncsrGetNumberOfRows(csrEV) =", csrGetNumberOfRows(csrEV)
print "\n>>> csrGetNumberOfColumns"
print "\ncsrGetNumberOfColumns(csrFV) =", csrGetNumberOfColumns(csrFV)
print "\ncsrGetNumberOfColumns(csrEV) =", csrGetNumberOfColumns(csrEV)

```

Macro referenced in 33b.

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.

```

⟨Sparse to dense matrix transformation 6a⟩ ≡
def csr2DenseMatrix(CSRm):
    nrows = csrGetNumberOfRows(CSRm)
    ncolumns = csrGetNumberOfColumns(CSRm)
    ScipyMat = zeros((nrows,ncolumns),int)
    C = CSRm.tocoo()
    for triple in zip(C.row,C.col,C.data):
        ScipyMat[triple[0],triple[1]] = triple[2]
    return ScipyMat

```

Macro referenced in 33a.

```

⟨Test examples of Sparse to dense matrix transformation 6b⟩ ≡
print "\n>>> csr2DenseMatrix"
print "\nFV =\n", csr2DenseMatrix(csrFV)
print "\nEV =\n", csr2DenseMatrix(csrEV)

```

Macro referenced in 7, 33b.

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 33a.

2.2 Characteristic matrices

We define as *characteristic matrices* M_k ($0 \leq k \leq d$) 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

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

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 ≡
import sys; sys.path.insert(0, 'lib/py/')
from larcc 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.*

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

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 .

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

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



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 33a.

⟨ Test example of Matrix filtering to produce the boundary matrix 9a ⟩ ≡

```
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 33b.

⟨ Matrix filtering via a generic predicate 9b ⟩ ≡

```
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 33a.

⟨ Test example of Matrix filtering via a generic predicate 9c ⟩ ≡

```
print "\n>>> csrPredFilter"
CSRm = csrPredFilter(matrixProduct(csrFV, csrTranspose(csrEV)).T, GE(2)).T
print "\nccsrPredFilter(csrFE) =\n", csr2DenseMatrix(CSRm)
```

◇

Macro referenced in 33b.

2.3 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.

⟨Extraction of facets of a cell complex 10a⟩ ≡

```
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):
                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 [33a](#).

⟨Computation of cell adjacencies 10b⟩ ≡

```
def larCellAdjacencies(CSRm):
    CSRm = matrixProduct(CSRm,csrTranspose(CSRm))
    return CSRm
```

◇

Macro referenced in [33a](#).

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.

```

⟨Test examples of Extraction of facets of a cell complex 11a⟩ ≡
    """ 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)(MKPOLLS((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)(MKPOLLS((V,EV))))
    ◇

```

Macro referenced in [33b](#).

Visualization of cell numbers The adjacency 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 11b⟩ ≡

    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 [33b](#).

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.

⟨ Visualization of cell indices 12a ⟩ ≡

```
""" Visualization of cell indices """
from sysml 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 12ab.

Macro referenced in 33a.

⟨ Visualization of cell indices 12b ⟩ ≡

```
""" Numbered visualization of a LAR model """
def larModelNumbering(scalx=1,scaly=1,scalz=1):
    def larModelNumbering0(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
```

◇

Macro defined by 12ab.
Macro referenced in 33a.

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.

```

⟨Drawing of oriented edges 13a⟩ ≡
    """ 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
    ◇

```

Macro referenced in 33a.

Example of oriented edge drawing An example of drawing of oriented edges is given in test/py/larcc/test11.py file, and in Figure 3, showing both the numbering of the cells and the arrows indicating the edge orientation is illustrated in Figure 3, where also the oriented boundary is shown.

```

"test/py/larcc/test11.py" 13b ≡
    """ Example of oriented edge drawing """
    import sys;sys.path.insert(0, 'lib/py/')
    from larcc import *

    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 3: 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)(MKPOLLS((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)
◇

```

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 \mathbf{C} , \mathbf{F} , \mathbf{E} , and \mathbf{V} , the 3-cells and their faces, edges and vertices, respectively. The input is the full-fledged LAR representation provided by

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

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

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

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

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

$$\mathbf{CF} := \mathbf{CV} \times \mathbf{FV}^t = \text{CSR}(M_3) \times \text{CSR}(M_2)^t \tag{5}$$

$$\mathbf{CE} := \mathbf{CV} \times \mathbf{EV}^t = \text{CSR}(M_3) \times \text{CSR}(M_1)^t \tag{6}$$

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

The other possible operators follow from a similar 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 16a⟩ ≡
    """ 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 16b⟩
    ⟨Cell-Edge incidence operator 16c⟩
    ⟨Face-Edge incidence operator 17a⟩
    ◇

```

Macro referenced in 33a.

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

```

⟨Cell-Face incidence operator 16b⟩ ≡
    """ Cell-Face incidence operator """
    def csrCellFaceIncidence(CV,FV):
        return boundary(FV,CV)

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

```

Macro referenced in 16a.

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

```

⟨Cell-Edge incidence operator 16c⟩ ≡
    """ Cell-Edge incidence operator """
    def csrCellEdgeIncidence(CV,EV):
        return boundary(EV,CV)

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

```

Macro referenced in 16a.

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

```

⟨Face-Edge incidence operator 17a⟩ ≡
    """ Face-Edge incidence operator """
    def csrFaceEdgeIncidence(FV,EV):
        return boundary(EV,FV)

    def larFaceEdge(FV,EV):
        return larIncidence(FV,EV)
    ◇

```

Macro referenced in [16a](#).

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" 17b ≡
    """ A mesh model and various incidence operators """
    import sys; sys.path.insert(0, 'lib/py/')
    from larcc import *
    from largrid 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 ($3 \leq k \leq 0$) 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:

$$\text{CF}_d = M_{d-1} M_d^t,$$

with

$$\text{CF}_d := \{a_{ij}\}, \quad a_{ij} = \begin{cases} 1 & \text{if } M_{d-1}(i)M_d(j) = |f_j| \\ 0 & \text{otherwise} \end{cases}$$

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 $[VV, EV, FV, CV, \dots]$. Notice that the function returns the inverse sequence $[EV, FE, CF, \dots]$, i.e., CF_k ($1 \leq k \leq d$).

```

⟨Incidence chain computation 18a⟩ ≡
    """ 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 [33a](#).

```

"test/py/larcc/test13.py" 18b ≡
    """ Example of incidence chain computation """
    import sys; sys.path.insert(0, 'lib/py/')
    from larcc import *
    from largrid 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 [4](#) for this purpose.

```

⟨Incidence chain for a 3D cuboidal complex 18c⟩ ≡
    incidence CF = [[0,2,4,6,8,9],[1,3,5,7,9,10]]

    incidence FE = [[0,2,8,9],[1,3,9,10],[4,6,11,12],[5,7,12,13],[0,4,14,15],

```

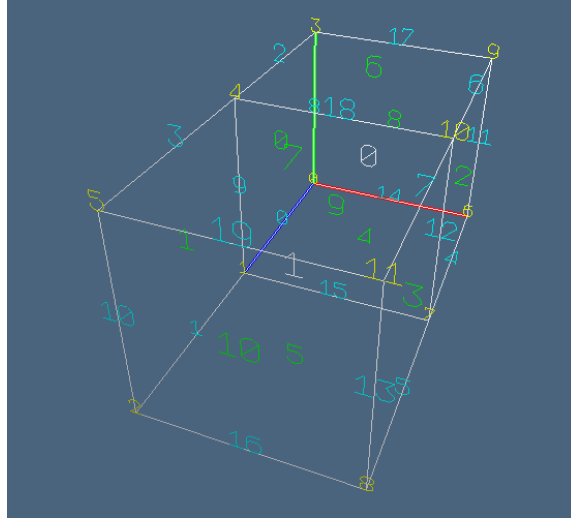


Figure 4: The stack of incidence relations gives the common links between cell numerations.

$[1, 5, 15, 16], [2, 6, 17, 18], [3, 7, 18, 19], [8, 11, 14, 17], [9, 12, 15, 18], [10, 13, 16, 19]]$

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.

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 $\mathbb{Z}_2 = \{0, 1\}$.

⟨ Test examples of From cells and facets to boundary operator 19 ⟩ ≡

V = $[[0.0, 0.0, 0.0], [1.0, 0.0, 0.0], [0.0, 1.0, 0.0], [1.0, 1.0, 0.0],$
 $[0.0, 0.0, 1.0], [1.0, 0.0, 1.0], [0.0, 1.0, 1.0], [1.0, 1.0, 1.0]]$

```

CV = [[0,1,2,4],[1,2,4,5],[2,4,5,6],[1,2,3,5],[2,3,5,6],[3,5,6,7]]
FV = [[0,1,2],[0,1,4],[0,2,4],[1,2,3],[1,2,4],[1,2,5],[1,3,5],[1,4,5],[2,3,5],
      [2,3,6],[2,4,5],[2,4,6],[2,5,6],[3,5,6],[3,5,7],[3,6,7],[4,5,6],[5,6,7]]
EV = [[0,1],[0,2],[0,4],[1,2],[1,3],[1,4],[1,5],[2,3],[2,4],[2,5],
      [2,6],[3,5],[3,6],[3,7],[4,5],[4,6],[5,6],[5,7],[6,7]]
VV = AA(LIST)(range(len(V)))

print "\ncoboundary_2 =\n", csr2DenseMatrix(coboundary(CV,FV))
print "\ncoboundary_1 =\n", csr2DenseMatrix(coboundary(FV,EV))
print "\ncoboundary_0 =\n", csr2DenseMatrix(coboundary(EV,VV))

```

Macro referenced in 33b.

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 20a ⟩ ≡

```

def boundary(cells,facets):
    lenV = max(max(cells),max(facets))
    csrCV = csrCreate(cells,lenV)
    csrFV = csrCreate(facets,lenV)
    csrFC = matrixProduct(csrFV, csrTranspose(csrCV))
    facetLengths = [csrCell.getnnz() for csrCell in csrCV]
    return csrBoundaryFilter(csrFC,facetLengths)

def coboundary(cells,facets):
    Boundary = boundary(cells,facets)
    return csrTranspose(Boundary)

```

Macro referenced in 33a.

⟨ From cells and facets to boundary cells 20b ⟩ ≡

```

def totalChain(cells):
    return csrCreate([[0] for cell in cells]) # ??? zero ??

def boundaryCells(cells,facets):
    csrBoundaryMat = boundary(cells,facets)
    csrChain = totalChain(cells)
    csrBoundaryChain = matrixProduct(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 33a.

```

⟨ Test examples of From cells and facets to boundary cells 21 ⟩ ≡
    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 [33b](#).

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 $FV = \text{CSR}(M_{d-1})$ and $CV = \text{CSR}(M_d)$, 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 $\langle v_0, \dots, v_{k-1}, v_{k+1}, \dots, v_d \rangle$ of the $\langle v_0, \dots, v_d \rangle$ cell.

\langle Signed boundary matrix for simplicial models 22a $\rangle \equiv$

```
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
    faceSigns = AA(C(POWER)(-1))(missingVertIndices)

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

◇

Macro referenced in [33a](#).

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 22b $\rangle \equiv$

```
def swap(mylist): return [mylist[1]]+[mylist[0]]+mylist[2:]

def boundaryCellsCocells(cells,facets):
```

```

csrSignedBoundaryMat = signedSimplicialBoundary(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

```

◇

Macro referenced in 33a.

Signed boundary matrix for polytopal complexes

⟨Signed boundary matrix for polytopal complexes 23⟩ ≡

```

""" 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 33a.

Oriented boundary cells for polytopal complexes

⟨Signed boundary cells for polytopal complexes 24a⟩ ≡

```

""" 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 33a.

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 integer vertices V of the generated complex, and the list of `bases` of cells of dimension k ($0 \leq k \leq d$), where $d = \text{len}(\text{shape}) - 1$.

```

"test/py/larcc/test14.py" 24b ≡
""" Boundary of a 2D cuboidal grid """
import sys;sys.path.insert(0, 'lib/py/')
from larcc import *

V,bases = larCuboids([6,6],True)
[VV,EV,FV] = bases
submodel = mkSignedEdges((V,EV))
VIEW(submodel)
VIEW(larModelNumbering(scalx=1,scaly=1,scalz=1)(V,bases,submodel,1))

orientedBoundary = signedCellularBoundaryCells(V,bases)
FV = [EV[e] if sign==1 else REVERSE(EV[e]) for (sign,e) in zip(*orientedBoundary)]
submodel = mkSignedEdges((V,FV))
VIEW(submodel)
VIEW(larModelNumbering(scalx=1,scaly=1,scalz=1)(V,bases,submodel,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" 25a ≡
    """ Oriented cuboidal and simplicial cells (same algorithm) """
    import sys;sys.path.insert(0, 'lib/py/')
    from larcc 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))))
    ◇
```

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

```
"test/py/larcc/test17.py" 25b ≡
    """ Boundary orientation of a random 2D cubical complex """
    import sys;sys.path.insert(0, 'lib/py/')
    from scipy import linalg
    from larcc 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

Top-down implementation

"test/py/larcc/test16.py" 26 ≡

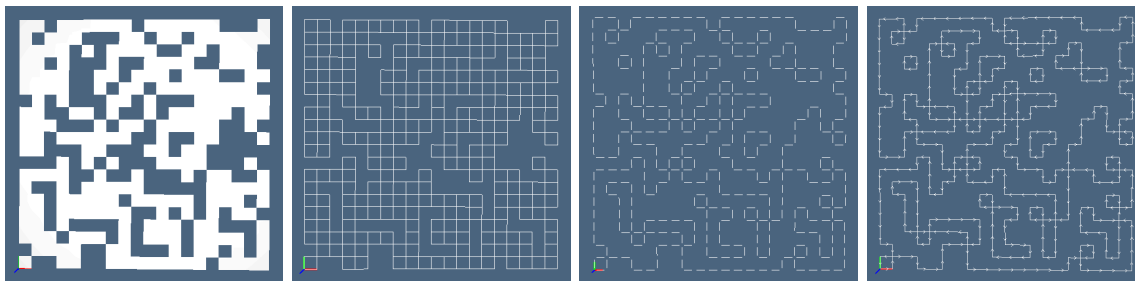


Figure 5: The orientation of the boundary of a random cuboidal 2-complex; (a) 2-cells; (b) 1-cells; (c) exploded boundary 1-chain; (d) oriented boundary 1-chain.

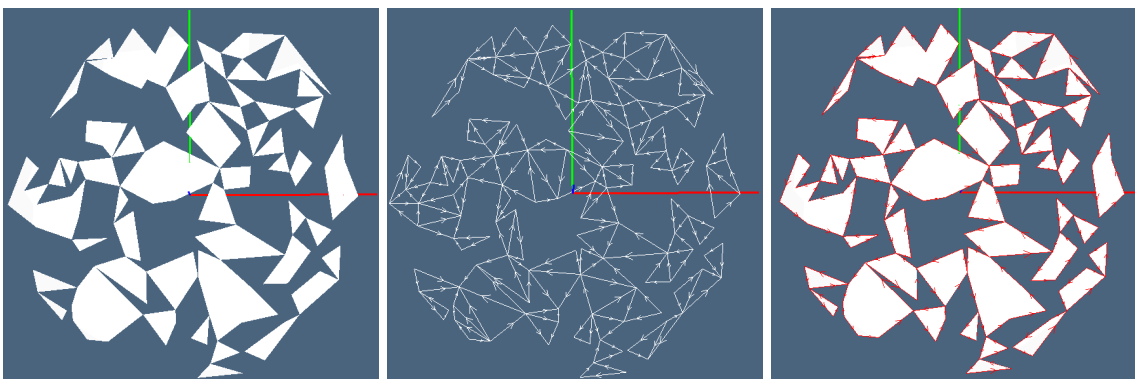


Figure 6: The orientation of the boundary of a random simplicial 2-complex; (a) 2-cells; (b) 1-cells; (c) oriented boundary 1-chain (red).

```

""" Boundary orientation of a random 2D triangulation """
import sys;sys.path.insert(0, 'lib/py/')
from scipy import linalg
from larcc import *
from random import random

⟨ Vertices V generated as random point in the unit circle 28a ⟩
⟨ Delaunay triangulation of the whole set V of points 28b ⟩
⟨ Fraction of triangles randomly discarded 29a ⟩
⟨ Coherently orient the input LAR model (V,FV) 29b ⟩
⟨ Compute the 1-cell and 0-cell bases EV and VV 29c ⟩
⟨ Signed 2-boundary matrix and signed boundary 1-chain 29d ⟩
⟨ Display the boundary 1-chain 30a ⟩
◇

```

Vertices V generated as random point in the unit circle

```

⟨ Vertices V generated as random point in the unit circle 28a ⟩ ≡
""" Vertices V generated as random point in the unit circle """
verts = []
npoints = 200
for k in range(npoints):
    t = 2*pi*random()
    u = random()+random()
    if u > 1: r = 2-u
    else: r = u
    verts += [[r*cos(t), r*sin(t)]]
VIEW(STRUCT(AA(MK)(verts)))
◇

```

Macro referenced in 26.

Delaunay triangulation of the whole set V of points

```

⟨ Delaunay triangulation of the whole set V of points 28b ⟩ ≡
""" Delaunay triangulation of the whole set V of points """
triangles = 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
◇

```

Macro referenced in 26.

Fraction of triangles randomly discarded

```
⟨ Fraction of triangles randomly discarded 29a ⟩ ≡
    """ Fraction of triangles randomly discarded """
    fraction = 0.7
    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 26.

Coherent orientation of input LAR model (V,FV)

```
⟨ Coherently orient the input LAR model (V,FV) 29b ⟩ ≡
    """ 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 26.

Compute the 1-cell and 0-cell bases EV and VV

```
⟨ Compute the 1-cell and 0-cell bases EV and VV 29c ⟩ ≡
    """ 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 26.

Signed boundary matrix $[\partial_2]$ and signed boundary 1-chain

```
⟨ Signed 2-boundary matrix and signed boundary 1-chain 29d ⟩ ≡
    """ 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 26.

Display the boundary 1-chain

```

⟨Display the boundary 1-chain 30a⟩ ≡
    """ Display the boundary 1-chain """
    VIEW(STRUCT(MKPOLS((V,FV))))
    VIEW(STRUCT(
        MKPOLS((V,FV)) +
        [COLOR(RED)(mkSignedEdges((V,cells))))])
    )
    ◇

```

Macro referenced in [26](#).

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; `Id` : identity matrix; `basis` : orthonormal spanning set of vectors e_k ; `vector` : position vector of a simplex vertex in translated coordinates; `unUsedIndices` : cell indices not moved to simplex.

```

⟨Oriented boundary cells for simplicial models 30b⟩ ≡
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 += [cell[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] for v in simplex] for simplex in simplices]
    return [SIGN(np.linalg.det(vcell)) for vcell in vcells]

```

◇

Macro referenced in [33a](#).

4 Exporting the library

4.1 MIT licence

⟨The MIT Licence 31⟩ ≡

```

"""
The MIT License
=====

```

```

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a copy of this software and associated documentation files (the
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without limitation the rights to use, copy, modify, merge, publish,
distribute, sublicense, and/or sell copies of the Software, and to
permit persons to whom the Software is furnished to do so, subject to

```


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"""

◇

Macro referenced in 33a.

4.2 Importing of modules or packages

⟨Importing of modules or packages 32⟩ ≡

```
from pyplasm import *
import collections
import scipy
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

from lar2psm import *
from sysml import *
◇
```

Macro referenced in 33a.

4.3 Writing the library file

```
"lib/py/larcc.py" 33a ≡
# -*- coding: utf-8 -*-
""" Basic LARCC library """
<The MIT Licence 31>
<Importing of modules or packages 32>
<From list of triples to scipy.sparse 3b>
<Brc to Coor 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>
<From cells and facets to boundary operator 20a>
<From cells and facets to boundary cells 20b>
<Signed boundary matrix for simplicial models 22a>
<Orientation of general convex cells 22b>
<Computation of cell adjacencies 10b>
<Extraction of facets of a cell complex 10a>
<Some incidence operators 16a>
<Visualization of cell indices 12a, ... >
<Numbered visualization of a LAR model ?>
<Drawing of oriented edges 13a>
<Incidence chain computation 18a>
<Signed boundary matrix for polytopal complexes 23>
<Signed boundary cells for polytopal complexes 24a>
<Oriented boundary cells for simplicial models 30b>

if __name__ == "__main__":
    <Test examples 33b>
    ◊
```

5 Unit tests

<Test examples 33b> ≡

```
<Test example of Brc to Coor transformation 3a>
<Test example of Coor 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>
<Test example of Matrix filtering via a generic predicate 9c>
```

{ Test examples of From cells and facets to boundary operator 19 }
 { Test examples of From cells and facets to boundary cells 21 }
 { Test examples of Computation of cell adjacencies 11b }
 { Test examples of Extraction of facets of a cell complex 11a }
 ◇

Macro referenced in 33a.

Comparing oriented and unoriented boundary

"test/py/larcc/test09.py" 34a ≡

```

""" 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" 34b ≡

```

""" 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" 35a ≡
import sys; sys.path.insert(0, 'lib/py/')
from larcc import *
from largrid import *
⟨input of 2D topology and geometry data 35b⟩
⟨characteristic matrices 35c⟩
⟨incidence matrix 35d⟩
⟨boundary and coboundary operators 36a⟩
⟨product of cell complexes 36b⟩
⟨2-skeleton extraction 36c⟩
⟨1-skeleton extraction 36d⟩
⟨0-coboundary computation 37a⟩
⟨1-coboundary computation 37b⟩
⟨2-coboundary computation 37c⟩
⟨boundary chain visualisation 37d⟩
◇
```

⟨input of 2D topology and geometry data 35b⟩ ≡

```
# 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 35a.

```
⟨characteristic matrices 35c⟩ ≡
# characteristic matrices
csrFV = csrCreate(FV)
csrEV = csrCreate(EV)
print "\nFV =\n", csr2DenseMatrix(csrFV)
print "\nEV =\n", csr2DenseMatrix(csrEV)
◇
```

Macro referenced in 35a.

```
⟨incidence matrix 35d⟩ ≡
# product
csrEF = matrixProduct(csrEV, csrTranspose(csrFV))
print "\nEF =\n", csr2DenseMatrix(csrEF)
◇
```

Macro referenced in 35a.

```

⟨boundary and coboundary operators 36a⟩ ≡
    # 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 [35a](#).

```

⟨product of cell complexes 36b⟩ ≡
    # product operator
    mod_2D = (V2,FV)
    V1,topol_0 = [[0.],[1.],[2.]], [[0],[1],[2]]
    topol_1 = [[0,1],[1,2]]
    mod_0D = (V1,topol_0)
    mod_1D = (V1,topol_1)
    V3,CV = larModelProduct([mod_2D,mod_1D])
    mod_3D = (V3,CV)
    VIEW(EXPLODE(1.2,1.2,1.2)(MKPOLLS(mod_3D)))
    print "\nk_3 =", len(CV), "\n"
    ◇

```

Macro referenced in [35a](#).

```

⟨2-skeleton extraction 36c⟩ ≡
    # 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)(MKPOLLS(SK2)))
    print "\nk_2 =", len(FV3), "\n"
    ◇

```

Macro referenced in [35a](#).

```

⟨1-skeleton extraction 36d⟩ ≡
    # 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)(MKPOLSK1)))
print "\nk_1 =", len(EV3), "\n"
◇

```

Macro referenced in 35a.

```

⟨0-coboundary computation 37a⟩ ≡
# 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 35a.

```

⟨1-coboundary computation 37b⟩ ≡
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 35a.

```

⟨2-coboundary computation 37c⟩ ≡
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 35a.

```

⟨boundary chain visualisation 37d⟩ ≡
# 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)(MKPOLSK1)))
◇

```

Macro referenced in 35a.

A.2 Boundary of 3D simplicial grid

```
"test/py/larcc/test02.py" 38a ≡
    import sys; sys.path.insert(0, 'lib/py/')
    ⟨boundary of 3D simplicial grid 38b⟩
    ◇

⟨boundary of 3D simplicial grid 38b⟩ ≡
    from simplexn import *
    from larcc import *

    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 38a.

A.3 Oriented boundary of a random simplicial complex

```
"test/py/larcc/test03.py" 38c ≡
    ⟨Importing external modules 38d⟩
    ⟨Generating and viewing a random 3D simplicial complex 39a⟩
    ⟨Computing and viewing its non-oriented boundary 39b⟩
    ⟨Computing and viewing its oriented boundary 39c⟩
    ◇
```

```
⟨Importing external modules 38d⟩ ≡
    import sys; sys.path.insert(0, 'lib/py/')
    from simplexn import *
    from larcc import *
```

```

from scipy import *
from scipy.spatial import Delaunay
import numpy as np
◇

```

Macro referenced in 38c.

```

⟨ Generating and viewing a random 3D simplicial complex 39a ⟩ ≡
    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]
    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[cell[2]][2]<0) and (verts[cell[3]][2]<0) ) ]
    V, CV = verts, cells
    VIEW(MKPOL([V,AA(AA(lambda k:k+1))(CV),[]]))
◇

```

Macro referenced in 38c.

```

⟨ Computing and viewing its non-oriented boundary 39b ⟩ ≡
    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)(MKPOL(bndry)))
◇

```

Macro referenced in 38c.

```

⟨ Computing and viewing its oriented boundary 39c ⟩ ≡
    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)(MKPOL(boundaryModel)))
◇

```

Macro referenced in 38c.

A.4 Oriented boundary of a simplicial grid

```

"test/py/larcc/test04.py" 39d ≡
    ⟨ Generate and view a 3D simplicial grid 40a ⟩
    ⟨ Computing and viewing the 2-skeleton of simplicial grid 40b ⟩
    ⟨ Computing and viewing the oriented boundary of simplicial grid 40c ⟩
◇

```



```

⟨ Generate and view a 3D simplicial grid 40a ⟩ ≡
    import sys; sys.path.insert(0, 'lib/py/')
    from simplexn import *
    from larcc import *
    V,CV = larSimplexGrid1([4,4,4])
    VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLs((V,CV))))
    ◇

```

Macro referenced in [39d](#).

```

⟨ Computing and viewing the 2-skeleton of simplicial grid 40b ⟩ ≡
    FV = larSimplexFacets(CV)
    EV = larSimplexFacets(FV)
    VIEW(EXPLODE(1.5,1.5,1.5)(MKPOLs((V,FV))))
    ◇

```

Macro referenced in [39d](#).

```

⟨ Computing and viewing the oriented boundary of simplicial grid 40c ⟩ ≡
    csrSignedBoundaryMat = signedSimplicialBoundary (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 [39d](#).

A.5 Skeletons and oriented boundary of a simplicial complex

```

"test/py/larcc/test05.py" 40d ≡
    import sys; sys.path.insert(0, 'lib/py/')

    ⟨ Skeletons computation and vilualisation 40e ⟩
    ⟨ Oriented boundary matrix visualization 41a ⟩
    ⟨ Computation of oriented boundary cells 41b ⟩
    ◇

```

```

⟨ Skeletons computation and vilualisation 40e ⟩ ≡
    from simplexn import *
    from larcc import *
    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 referenced in 40d.

```
< Oriented boundary matrix visualization 41a > ≡  
    np.set_printoptions(threshold='nan')  
    csrSignedBoundaryMat = signedSimplicialBoundary (FV,EV)  
    Z = csr2DenseMatrix(csrSignedBoundaryMat)  
    print "\ncsrSignedBoundaryMat =\n", Z  
    from pylab import *  
    matshow(Z)  
    show()  
◇
```

Macro referenced in 40d.

```
< Computation of oriented boundary cells 41b > ≡  
    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 40d.

A.6 Boundary of random 2D simplicial complex

```
"test/py/larcc/test06.py" 41c ≡  
    import sys; sys.path.insert(0, 'lib/py/')  
    from simplexn import *  
    from larcc import *  
    from scipy.spatial import Delaunay  
    < Test for quasi-equilateral triangles 41d >  
    < Generation and selection of random triangles 42a >  
    < Boundary computation and visualisation 42b >  
◇
```

```
< Test for quasi-equilateral triangles 41d > ≡  
    def quasiEquilateral(tria):  
        a = VECTNORM(VECTDIFF(tria[0:2]))  
        b = VECTNORM(VECTDIFF(tria[1:3]))  
        c = VECTNORM(VECTDIFF([tria[0],tria[2]]))  
        m = max(a,b,c)  
        if m/a < 1.7 and m/b < 1.7 and m/c < 1.7: return True  
        else: return False  
◇
```

Macro referenced in 41c.

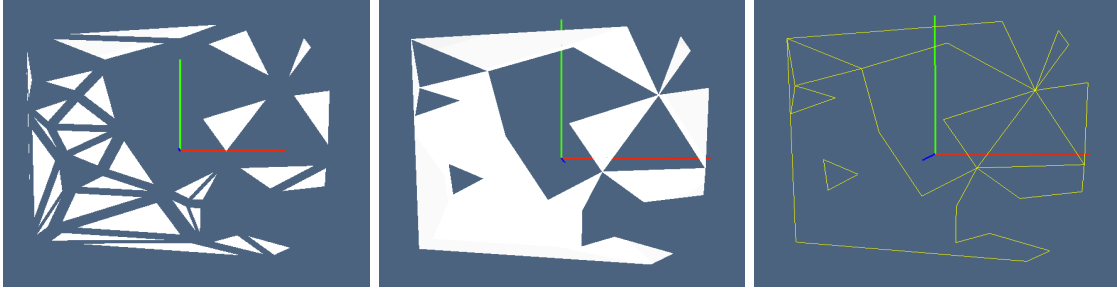


Figure 7: example caption

⟨ Generation and selection of random triangles 42a ⟩ ≡

```
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 [41c](#).

⟨ Boundary computation and visualisation 42b ⟩ ≡

```
orientedBoundary = signedBoundaryCells(V,FV,EV)
submodel = mkSignedEdges((V,orientedBoundary))
VIEW(submodel)
◇
```

Macro referenced in [41c](#).

⟨ Decompose a permutation into cycles 42c ⟩ ≡

```
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 permutation0rbits([2, 3, 4, 5, 6, 7, 0, 1])
    print [3,9,8,4,10,7,2,11,6,0,1,5]
    print permutation0rbits([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" 43a ≡
    import sys; sys.path.insert(0, 'lib/py/')
    from simplexn import *
    from larcc import *
    from largrid import *
    ⟨ Definition of 1-dimensional LAR models 43b ⟩
    ⟨ Assembly generation of squares and triangles 43c ⟩
    ⟨ Assembly generation of cubes and tetrahedra 44 ⟩

```

◇

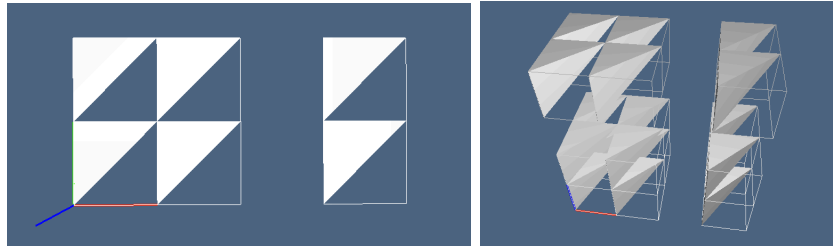


Figure 8: (a) Assemblies of squares and triangles; (b) assembly of cubes and tetrahedra.

```

⟨ Definition of 1-dimensional LAR models 43b ⟩ ≡
    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 43a.

```

⟨ Assembly generation of squares and triangles 43c ⟩ ≡
    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 43a.

⟨ Assembly generation of cubes and tetrahedra 44 ⟩ ≡

```
from largrid import *
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 43a.

References

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