

## 2023-06-19\_010\_Meshes\_Part-02

June 19, 2023

```
[1]: import numpy as np
import scipy
import imageio

import matplotlib
import matplotlib.pyplot as plt
import matplotlib.cm as cm

matplotlib.rc('image', interpolation='nearest')
matplotlib.rc('figure', facecolor='white')
matplotlib.rc('image', cmap='viridis')
colors=plt.rcParams['axes.prop_cycle'].by_key()['color']
%matplotlib inline

from matplotlib.animation import FuncAnimation
matplotlib.rc('animation', html='html5')
import colorcet as ccm
from graphplot import *
```

```
[2]: import FEM as FEM
```

### 1 Meshes: Regions and boundaries

- intuition: every surface has a boundary, which is a contour
- every path has a boundary, which consists of start and endpoint
- closed surfaces and cycles have an "empty" boundary
- this is the beginning of a whole branch of pure and abstract mathematics
- but now and then these questions arise in our "everyday research", then it is helpful to have heard about these ideas

## 1.1 Setup of a simple example mesh

```
[3]: # again: create/load a finer example mesh
data=np.load("triangulation.npz")
pointData=data["pointData"]
triangleData=data["triangleData"]
data.close()
nPoints=pointData.shape[0]
nTriangles=triangleData.shape[0]

# extract edge data
edgeData,etAdjacencyData,etAdjacencyDataOrientation=FEM.getEdges(triangleData)
nEdges=edgeData.shape[0]

[4]: def
    ↪drawEdges(ax,pointData,edgeData,colors="k",drawOrientations=False,arrowScale=0.
    ↪03,**kwargs):
        lineCollection=matplotlib.collections.
    ↪LineCollection(pointData[edgeData],color=colors,**kwargs)
        ax.add_collection(lineCollection)

        # draw small (manual) on edges
        if drawOrientations:
            edgeCoords=pointData[edgeData]
            midPoints=np.mean(edgeCoords,axis=1)
            orientations=edgeCoords[:,1,:]-edgeCoords[:,0,:]
            norms=np.linalg.norm(orientations,axis=1)
            orientations=np.einsum(orientations,[0,1],1./norms,[0],[0,1])
            for phi in [0.7*np.pi,-0.7*np.pi]:
                R=np.array([[np.cos(phi),-np.sin(phi)],[np.sin(phi),np.
    ↪cos(phi)]]),dtype=np.double)
                arr=np.einsum(R,[0,1],orientations,[2,1],[2,0])
                lines=np.zeros((nEdges,2,2))
                lines[:,0,:]=midPoints
                lines[:,1,:]=midPoints+arrowScale*arr
                lineCollection=matplotlib.collections.
    ↪LineCollection(lines,color=colors,**kwargs)
                ax.add_collection(lineCollection)

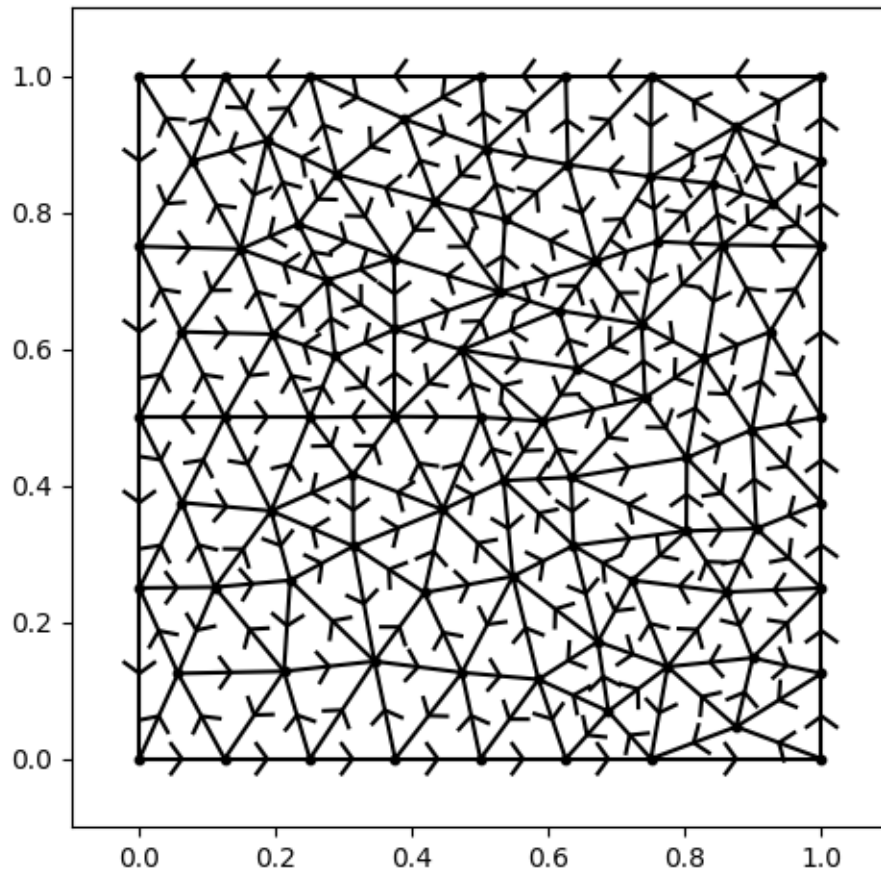
[5]: fig=plt.figure()
ax=fig.add_subplot(aspect=1.)
ax.scatter(pointData[:,0],pointData[:,1],c="k",s=10)

drawEdges(ax,pointData,edgeData,colors="k",drawOrientations=True,arrowScale=0.
    ↪03)
#ax.scatter(midPoints[:,0],midPoints[:,1])
```

```

buffer=0.1
plt.xlim([-buffer,1+buffer])
plt.ylim([-buffer,1+buffer])
plt.tight_layout()
plt.show()

```



## 1.2 Representing regions

- now we can represent 2d regions that are given as unions of triangles
- we simply use an indicator vector: 1 if triangle is part of region, 0 otherwise

### 1.2.1 Example: single triangle

```
[6]: # allocate an empty region, then set some index to True
region=np.zeros((nTriangles,),dtype=bool)
region[100]=True
```

### 1.2.2 Example: a few more triangles

```
[9]: # allocate an empty region, then set some index to True
# find all triangles that touch a given vertex
region=(np.sum(triangleData==20,axis=1)>0)
```

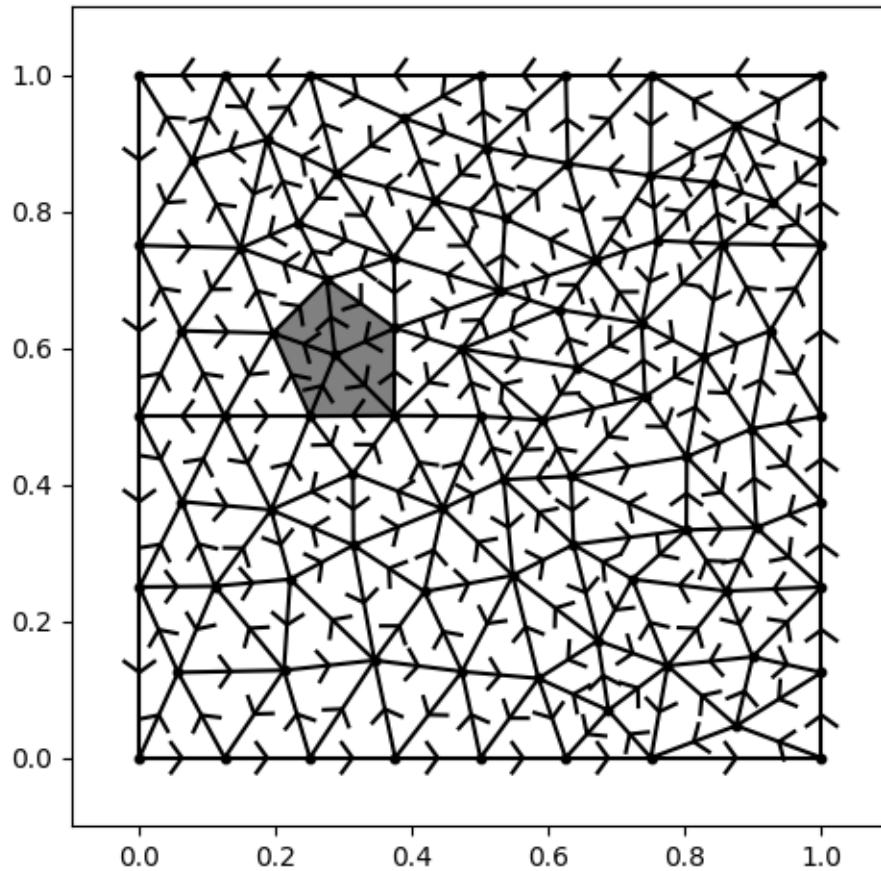
## 1.3 Visualization of region

```
[10]: # now visualize this as filled triangles
fig=plt.figure()
ax=fig.add_subplot(aspect=1.)
ax.scatter(pointData[:,0],pointData[:,1],c="k",s=10)

drawEdges(ax,pointData,edgeData,colors="k",drawOrientations=True)

polyCollection=matplotlib.collections.
    ↳LineCollection(pointData[triangleData[region]],lw=0,fc="#808080",zorder=-2)
ax.add_collection(polyCollection)

buffer=0.1
plt.xlim([-buffer,1+buffer])
plt.ylim([-buffer,1+buffer])
plt.tight_layout()
plt.show()
```



#### 1.4 Compute boundary of region

```
[12]: def getB2(nEdges,etAdjacencyData,etAdjacencyDataOrientation):
        nTriangles=etAdjacencyData.shape[0]
        data=etAdjacencyDataOrientation.copy().ravel()
        indices=etAdjacencyData.copy().ravel()
        indptr=3*np.arange(0,nTriangles+1)
        result=scipy.sparse.
        ↪csc_matrix((data,indices,indptr),shape=(nEdges,nTriangles)).tocsr()
        return result
```

```
[13]: # boundary operator: take indicator vector of triangles to indicator vector of ↪
        ↪bounding edges
        # take orientation into account: this way, contributions from adjacent ↪
        ↪triangles cancel

        # B2 stands for: boundary of 2-dimensional objects
```

```
B2=getB2(nEdges,etAdjacencyData,etAdjacencyDataOrientation)
```

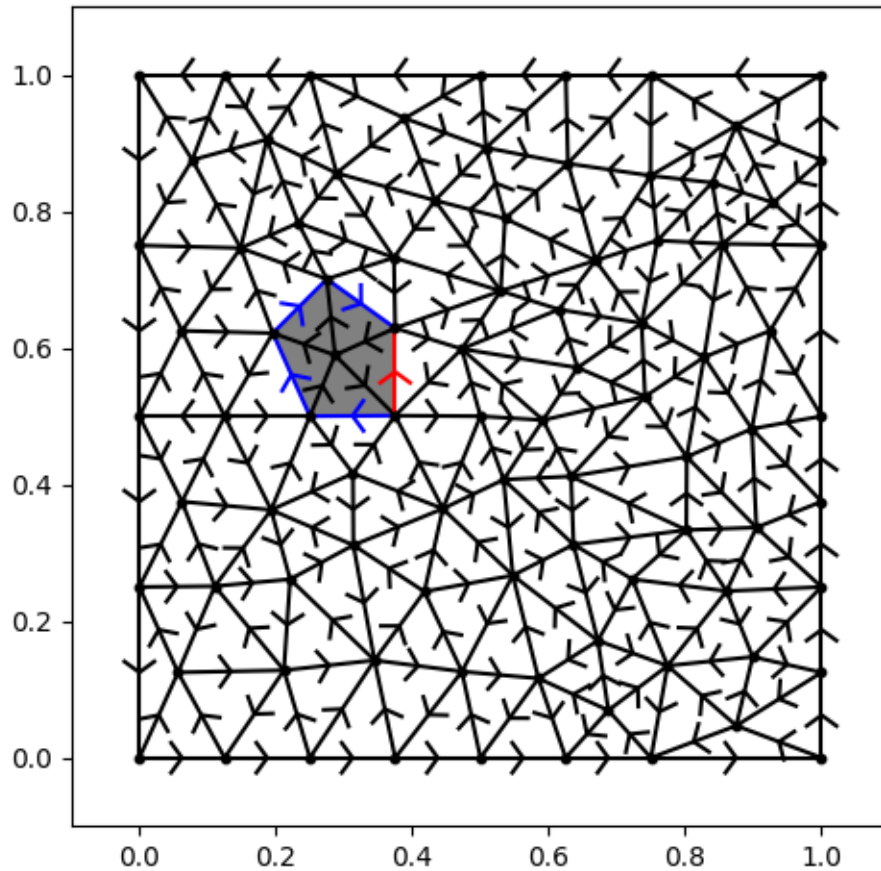
```
[14]: B2.shape
```

```
[14]: (237, 150)
```

```
[15]: # now apply to region indicator  
boundary=B2.dot(region)
```

```
[16]: def signToColor(bdry):  
        colors=np.zeros((bdry.shape[0],3))  
        colors[bdry>0.1]=np.array([1.,0.,0.])  
        colors[bdry<-0.1]=np.array([0.,0.,1.])  
        return colors
```

```
[17]: # highlight boundary in plot  
fig=plt.figure()  
ax=fig.add_subplot(aspect=1.)  
ax.scatter(pointData[:,0],pointData[:,1],c="k",s=10)  
  
lineColors=signToColor(boundary)  
drawEdges(ax,pointData,edgeData,colors=lineColors,drawOrientations=True,arrowScale=0.  
    ↪03,zorder=-1)  
  
polyCollection=matplotlib.collections.  
    ↪LineCollection(pointData[triangleData[region]],lw=0,fc="#808080",zorder=-2)  
ax.add_collection(polyCollection)  
  
buffer=0.1  
plt.xlim([-buffer,1+buffer])  
plt.ylim([-buffer,1+buffer])  
plt.tight_layout()  
plt.show()
```



## 1.5 Shortest paths

We can use Dijkstra's algorithm (or others) to find shortest paths on edges through the mesh

```
[18]: def getShortestPaths(pointData,edgeData):
    # compute edge lengths
    edgeLengths=np.linalg.norm(pointData[edgeData[:,0]]-pointData[edgeData[:,1]],axis=1)
    # build weighted adjacency matrix of graph, with weights given by lengths

    if False:
        # note: edges in the graph that we want to perform the search on, are
        # un-oriented, but the computational
        # representation has oriented edges. so we need to add both directions
        # the construction is most efficiently done in scipy.sparse.coo_matrix:
        graph=scipy.sparse.coo_matrix((\
            np.concatenate((edgeLengths,edgeLengths)),\
            (np.concatenate((edgeData[:,0],edgeData[:,1])),\
```

```

        np.concatenate((edgeData[:,1],edgeData[:,0])))\
    ))
    # non-existing entries correspond to infinite edge lengths (not zero!)
    # convert to csr, which is (probably) faster in graph algorithm
    graph=graph.tocsr()
    else:
        # use "directed=False" option of graph search instead
        graph=scipy.sparse.coo_matrix((edgeLengths,(edgeData[:,0],edgeData[
↪,1])))

    dist,pred=scipy.sparse.csgraph.
↪shortest_path(graph,directed=False,return_predecessors=True)
    return dist,pred

```

```

[19]: def extractPath(pred,i0,i1):
    path=[i1]
    j=i1
    while j!=i0:
        j=pred[i0,j]
        path.append(j)
    return np.array(path,dtype=np.int32)

def translatePath(path,edgeData):
    nEdges=edgeData.shape[0]
    # these matrices will help identifying edge indices from the vertex indices:
    auxmatInd=scipy.sparse.coo_matrix((\
        np.concatenate((np.arange(nEdges),np.arange(nEdges))),\
        (np.concatenate((edgeData[:,0],edgeData[:,1])),\
        np.concatenate((edgeData[:,1],edgeData[:,0]))))\
    )).tocsr()
    auxmatOrientation=scipy.sparse.coo_matrix((\
        np.concatenate((np.full((nEdges,),fill_value=1.),np.
↪full((nEdges,),fill_value=-1.))),\
        (np.concatenate((edgeData[:,0],edgeData[:,1])),\
        np.concatenate((edgeData[:,1],edgeData[:,0]))))\
    )).tocsr()

    result=np.zeros((nEdges,),dtype=np.double)
    for j in range(len(path)-1,0,-1):
        ↵
↪result[auxmatInd[path[j],path[j-1]]]=auxmatOrientation[path[j],path[j-1]]
    return result

```



```

[20]: # plot length from a given vertex as color coded mesh
      # which vertex?
      i0=0
      # find shortest path to a given target vertex
      i1=30

      dist,pred=getShortestPaths(pointData,edgeData)
      pathList=extractPath(pred,i0,i1)
      path=translatePath(pathList,edgeData)

      fig=plt.figure(figsize=(6,6))
      ax=fig.add_subplot(aspect=1.)

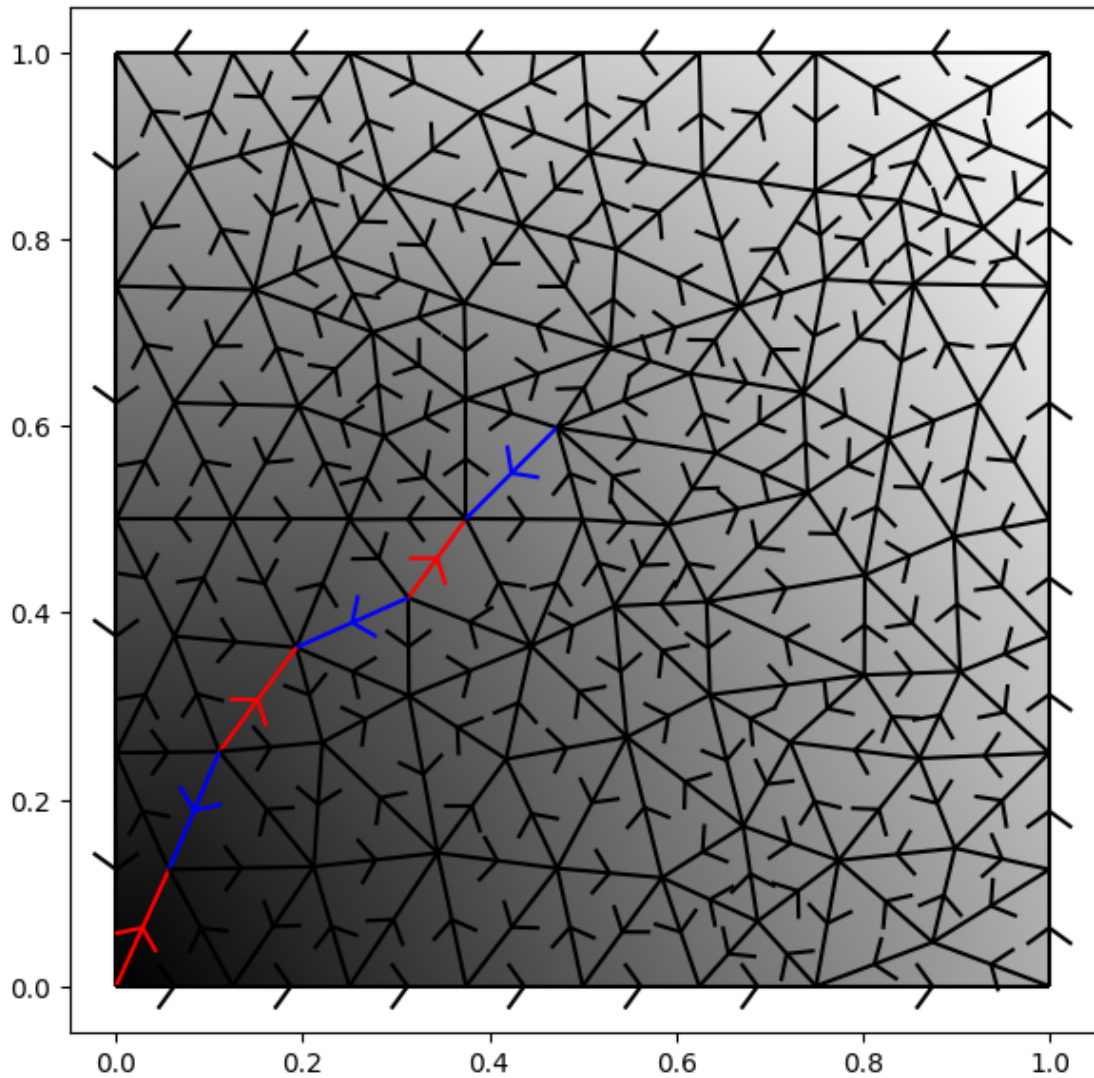
      tri=matplotlib.tri.Triangulation(pointData[:,0],pointData[:,1],triangleData)
      pltobj=ax.tripcolor(tri,dist[i0,:]/np.max(dist[i0,:]),shading='gouraud',cmap=cm.
          ↪gray)

      lineColors=signToColor(path)
      drawEdges(ax,pointData,edgeData,colors=lineColors,drawOrientations=True,arrowScale=0.
          ↪03,zorder=1)

      #plt.colorbar(pltobj)

      plt.tight_layout()
      plt.show()

```



## 1.6 Boundaries of paths

- We can interpret starting and end points of lines (paths) as their boundaries, weighted with opposite signs.
- A path from point A to point B will then have A as positive boundary and B as negative boundary.

```
[20]: def getB1(nPoints,edgeData):
      nEdges=edgeData.shape[0]
      data=np.zeros((nEdges*2),dtype=np.double)
      data[0::2]=1.
      data[1::2]=-1.
```

```

    indices=edgeData.copy().ravel()
    indptr=2*np.arange(nEdges+1)
    result=scipy.sparse.
↪csc_matrix((data,indices,indptr),shape=(nPoints,nEdges)).tocsr()
    return result

```

```
[21]: B1=getB1(nPoints,edgeData)
```

```

[22]: # starting vertex
i0=10
# end vertex
i1=53

pathList=extractPath(pred,i0,i1)
path=translatePath(pathList,edgeData)

fig=plt.figure(figsize=(6,6))
ax=fig.add_subplot(aspect=1.)

tri=matplotlib.tri.Triangulation(pointData[:,0],pointData[:,1],triangleData)
ax.tripcolor(tri,dist[i0,:]/np.max(dist[i0,:]),shading='gouraud',cmap=cm.gray)

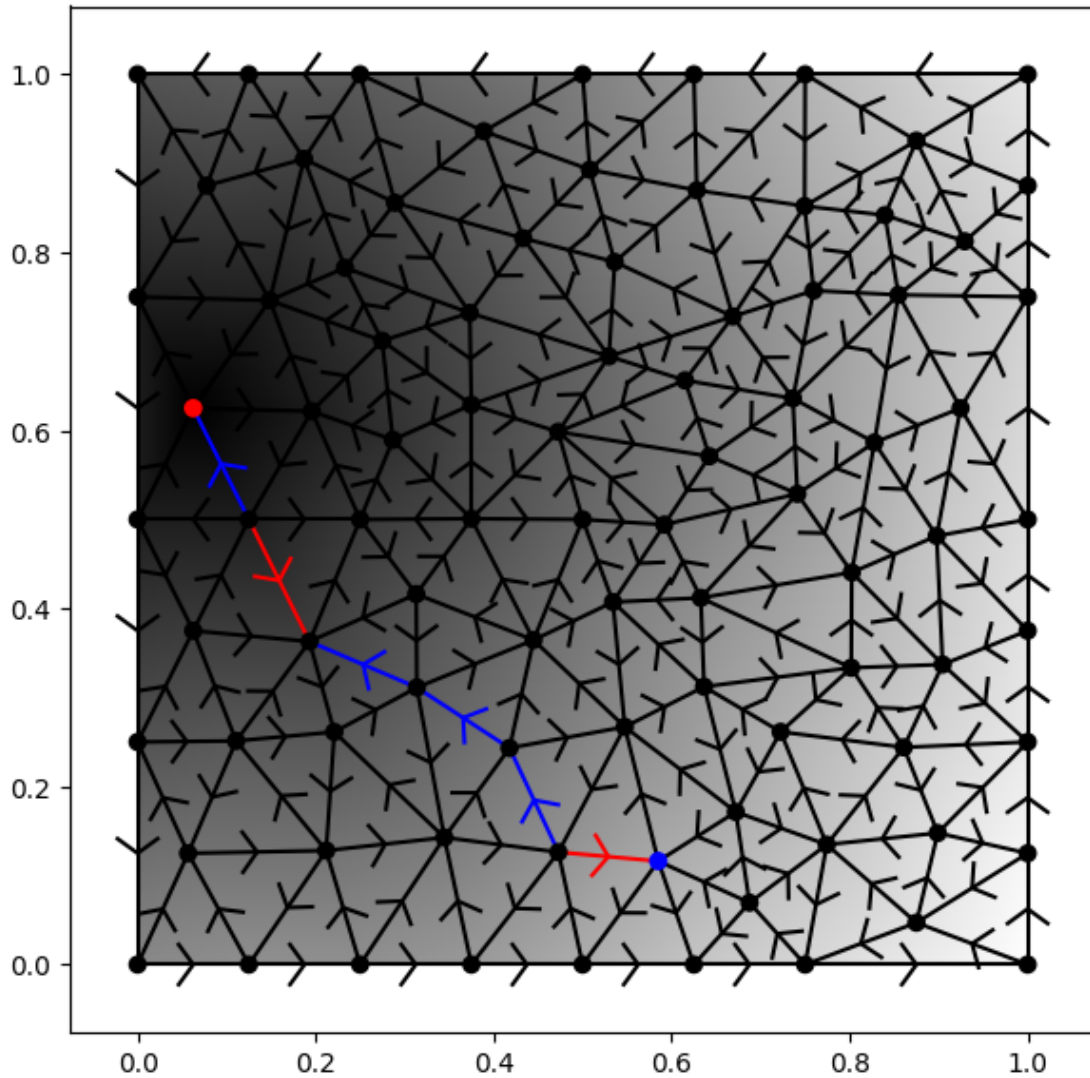
lineColors=signToColor(path)
drawEdges(ax,pointData,edgeData,colors=lineColors,drawOrientations=True,arrowScale=0.
↪03,zorder=1)

# add colored points to indicate boundary of shortest path
pathBoundary=B1.dot(path)
pointColors=signToColor(pathBoundary)
plt.scatter(pointData[:,0],pointData[:,1],c=pointColors,zorder=2)

#for i,x in enumerate(pointData):
#    ax.text(x[0],x[1],i,c="r")

plt.tight_layout()
plt.show()

```



## 1.7 Region from boundary

```
[23]: # first build a cycle (which we conjecture to be a boundary)
# fix some vertices, connect shortest paths between them
iList=[10,53,31,56,10]
path=np.zeros(nEdges)
for j in range(len(iList)-1):
    pathList=extractPath(pred,iList[j],iList[j+1])
    path+=translatePath(pathList,edgeData)

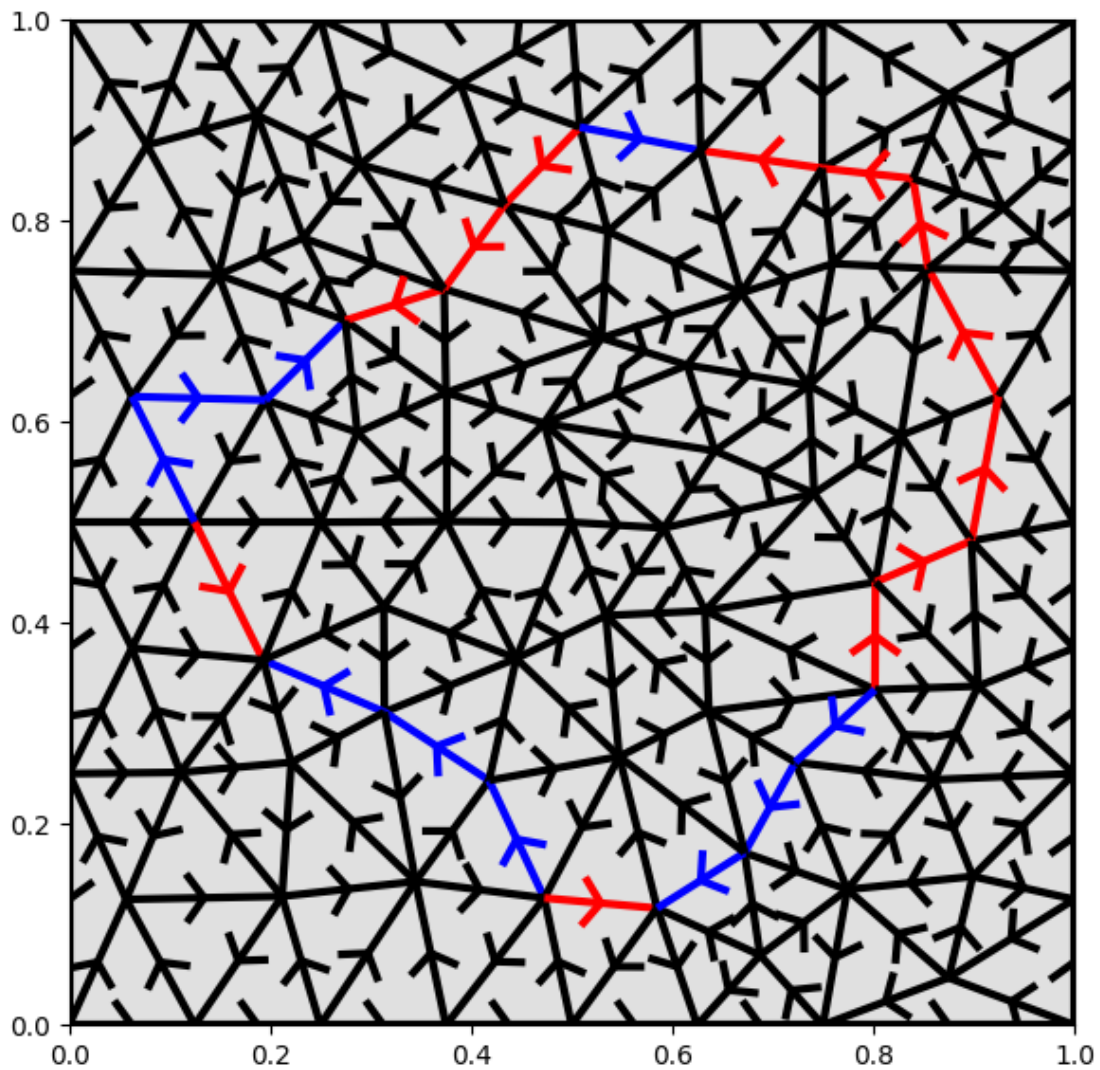
fig=plt.figure(figsize=(6,6),facecolor="w")
ax=fig.add_subplot(aspect=1.,facecolor="#e0e0e0")
```

```

lineColors=signToColor(path)
drawEdges(ax,pointData,edgeData,colors=lineColors,drawOrientations=True,arrowScale=0.
    ↪03,zorder=1,lw=3)

plt.tight_layout()
plt.show()

```



```
[24]: import scipy.sparse.linalg
```

```

[25]: # next: search for region indicator, such that it has the specified path as
    ↪boundary.
    # solve: B2.dot(region)=path

```

```

# since we do not know for sure if solution exists, use leastsquares ansatz
res=scipy.sparse.linalg.lsqr(B2,path)
region=res[0]
# test solution quality:
print("norm of discrepancy between B2(region) and path:",res[3])
print("if solution was found, this should be 1: ",res[1])

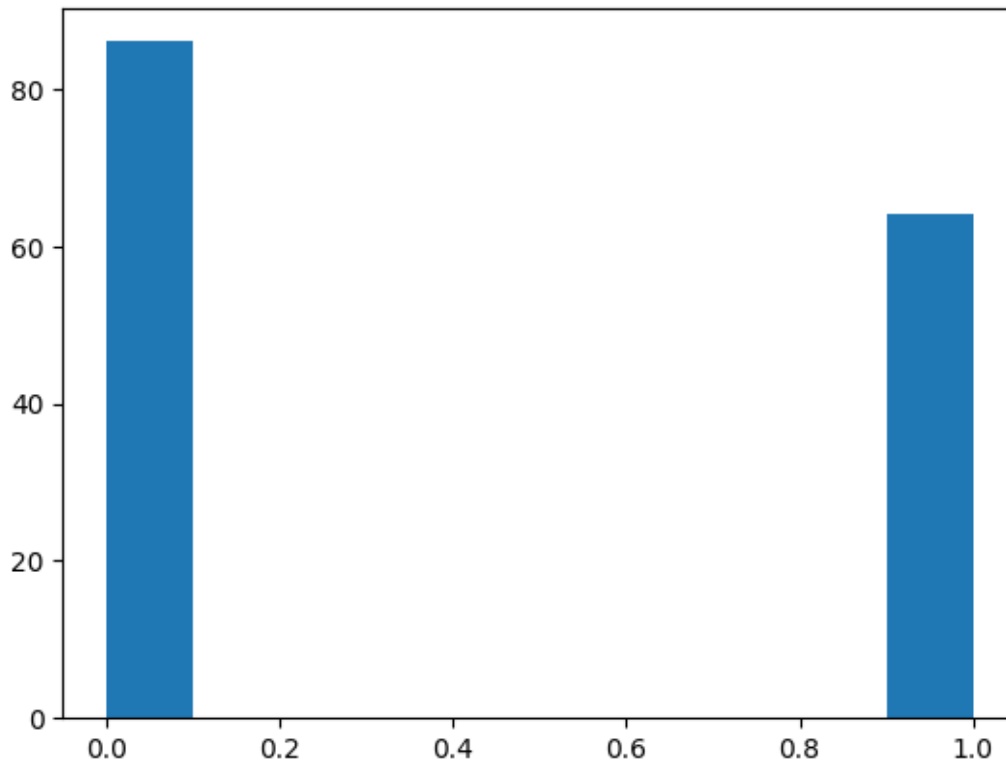
```

norm of discrepancy between B2(region) and path: 8.613445229696281e-05  
if solution was found, this should be 1: 1

```

[26]: # if a good solution was found, then region should be binary, with only 0 and 1
      ↪ entries
plt.hist(region)
plt.show()

```



```

[27]: # first build a cycle (which we conjecture to be a boundary)
      # fix three vertices, connect shortest paths between them

fig=plt.figure(figsize=(6,6),facecolor="w")
ax=fig.add_subplot(aspect=1.,facecolor="#e0e0e0")

lineColors=signToColor(path)

```

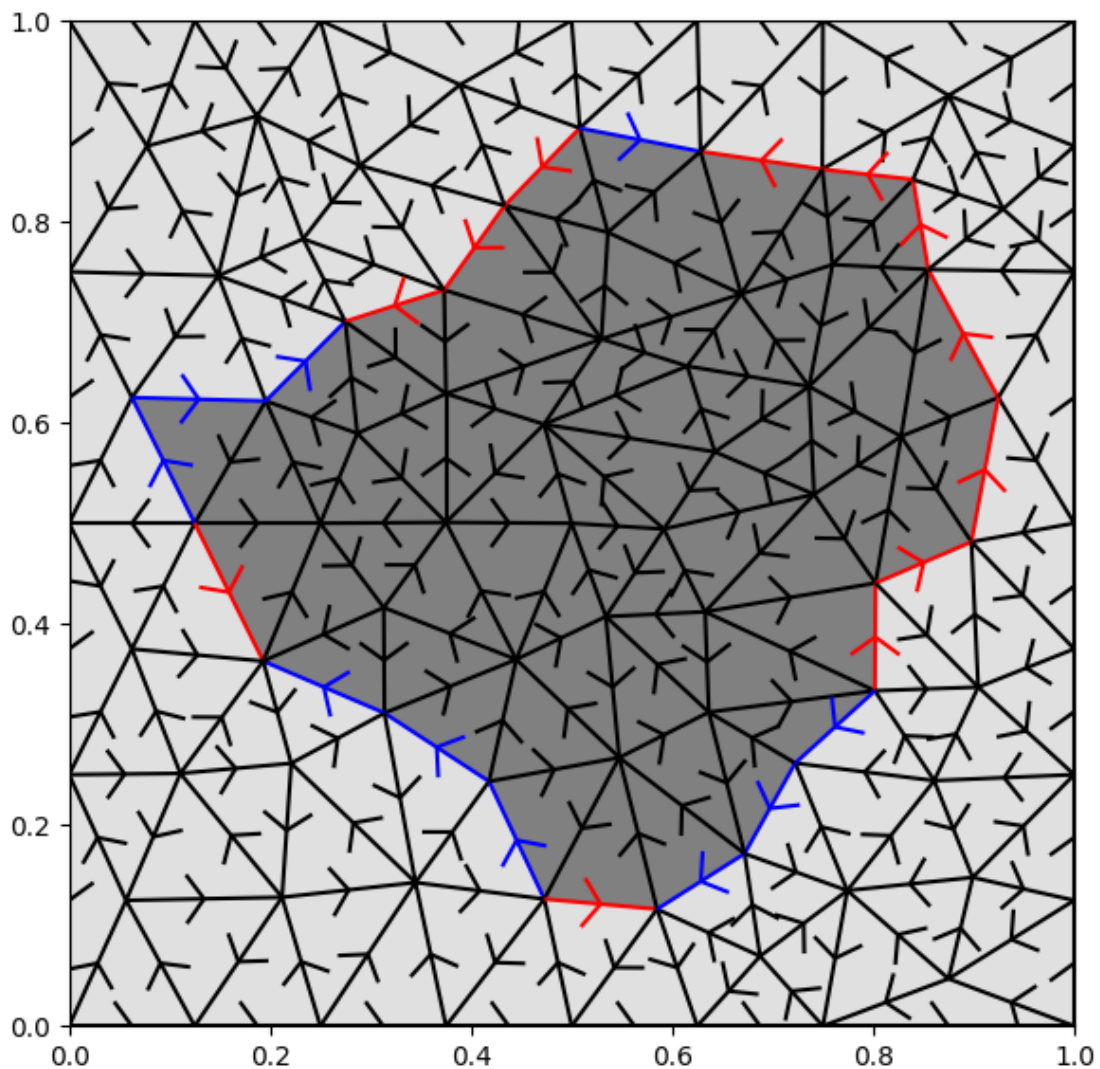
```

drawEdges(ax,pointData,edgeData,colors=lineColors,drawOrientations=True,arrowScale=0.
    ↪03,zorder=1)

regionIndicator=(region>0.1)
polyCollection=matplotlib.collections.
    ↪LineCollection(pointData[triangleData[regionIndicator]],\
        lw=0,fc="#808080",zorder=-2)
ax.add_collection(polyCollection)

plt.tight_layout()
plt.show()

```





## 1.8 Is every cycle path a boundary of a region?

```
[28]: # first observation: each boundary path of a region has zero boundary itself
# (boundaries of regions are cycle paths; these have no beginning or end point)
# confirm by showing that B1.dot(B2)= zero matrix
B1.dot(B2)
```

```
[28]: <88x150 sparse matrix of type '<class 'numpy.float64'>'
      with 0 stored elements in Compressed Sparse Row format>
```

```
[29]: # so: image of B2 lives in kernel of B1
# compare dimensions:
#   dim(ker B1)=nEdges-rank(B1)
#   dim(img B2)=rank(B2)
```

```
[30]: # since we use small matrices: determine rank on dense matrices
B1Dense=B1.todense()
B2Dense=B2.todense()

print(nEdges-np.linalg.matrix_rank(B1Dense))
print(np.linalg.matrix_rank(B2Dense))
```

150

150

```
[31]: # both numbers are equal: hence every cycle must be a boundary on our current_
      ↪ mesh
```

## 1.9 A mesh with a hole

```
[33]: # create/load a finer example mesh
if False:
    import lib.Triangle as Triangle

    points=np.array([[0,0],[0,3],[3,0],[3,3],[1,1],[1,2],[2,1],[2,2]],dtype=np.
    ↪double)
    segments=np.array([[4,5],[5,7],[7,6],[6,4]],dtype=np.int32)
    segmentMarkers=np.zeros(segments.shape[0],dtype=np.int32)
    holes=np.array([[1.5,1.5]],dtype=np.double)
    maxArea=0.1
    pointData,triangleData=Triangle.
    ↪createMeshAdvanced(points,segments,segmentMarkers,holes,\
        flagStr="zqpc",maxArea=maxArea)
    np.savez("triangulation_hole.
    ↪npz",pointData=pointData,triangleData=triangleData)
else:
```



```

data=np.load("triangulation_hole.npz")
pointData=data["pointData"]
triangleData=data["triangleData"]
data.close()
nPoints=pointData.shape[0]
nTriangles=triangleData.shape[0]

# extract edge data
edgeData,etAdjacencyData,etAdjacencyDataOrientation=FEM.getEdges(triangleData)
nEdges=edgeData.shape[0]

```

```

[34]: fig=plt.figure(figsize=(6,6),facecolor="w")
ax=fig.add_subplot(aspect=1.)

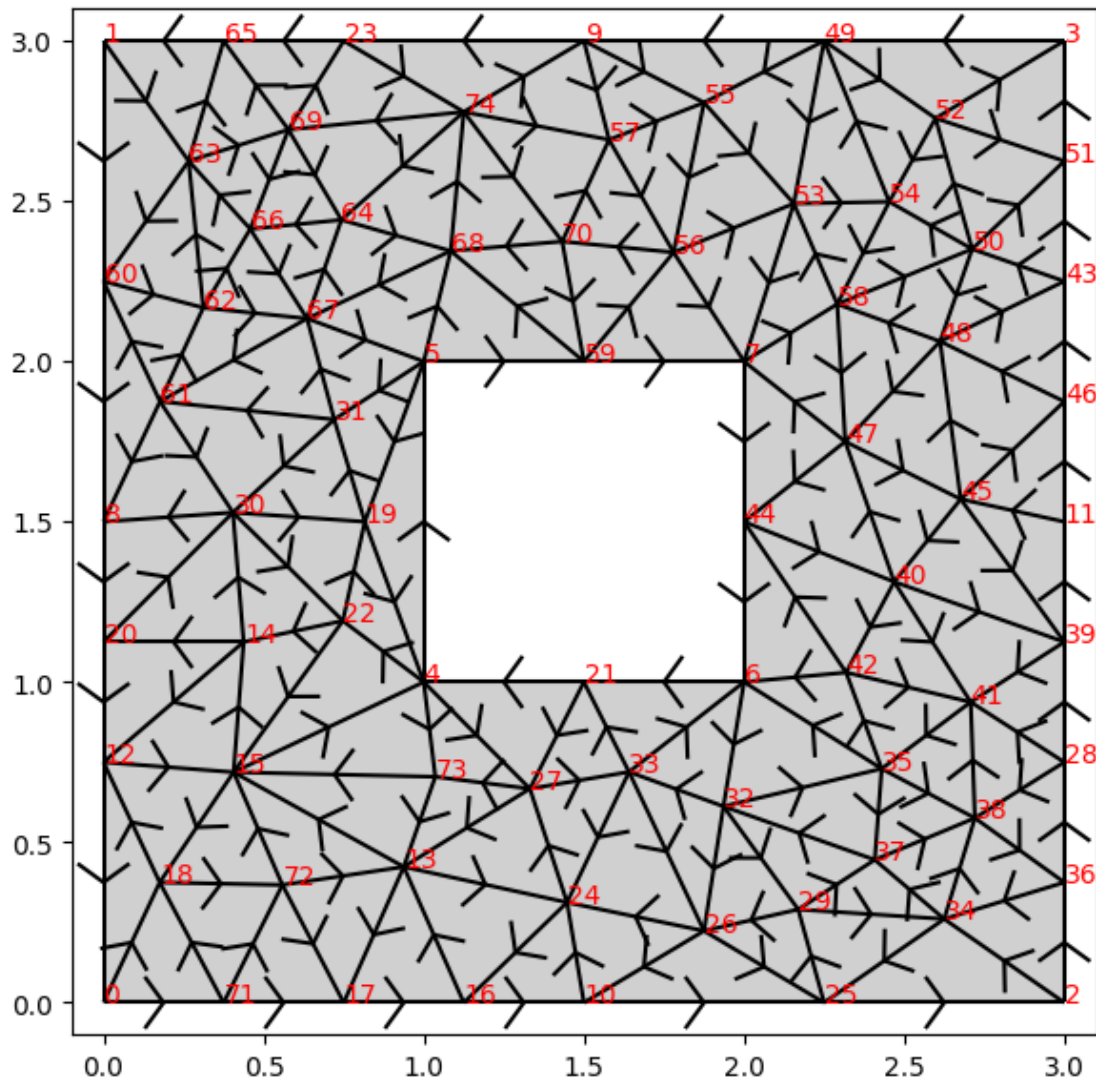
polyCollection=matplotlib.collections.
    ↳LineCollection(pointData[triangleData],lw=0,fc="#d0d0d0",zorder=-2)
ax.add_collection(polyCollection)
drawEdges(ax,pointData,edgeData,colors="k",drawOrientations=True,arrowScale=0.1)

for i,x in enumerate(pointData):
    ax.text(x[0],x[1],i,c="r")

# plot range
buffer=0.1
lim1,lim2,lim3,lim4=np.min(pointData[:,0]),np.max(pointData[:,0]),np.
    ↳min(pointData[:,1]),np.max(pointData[:,1])
plt.xlim([lim1-buffer,lim2+buffer])
plt.ylim([lim3-buffer,lim4+buffer])

plt.tight_layout()
plt.show()

```



```
[35]: B1=getB1(nPoints,edgeData)
      B2=getB2(nEdges,etAdjacencyData,etAdjacencyDataOrientation)
```

```
[36]: # do the rank test again:

      # recall:
      # -image of B2 lives in kernel of B1
      # -compare dimensions:
      #   dim(ker B1)=nEdges-rank(B1)
      #   dim(img B2)=rank(B2)

      # since we use small matrices: determine rank on dense matrices
      B1Dense=B1.todense()
```

```

B2Dense=B2.todense()

print(nEdges-np.linalg.matrix_rank(B1Dense))
print(np.linalg.matrix_rank(B2Dense))

```

120  
119

```

[37]: # there is one equivalence class of cycles that is NOT a boundary!
      # intuitively clear: these are the paths that go around the hole

```

```

[38]: # construct such a path and examine it in more detail
      dist,pred=getShortestPaths(pointData,edgeData)

```

```

[39]: # first build a cycle (which we conjecture to be a boundary)
      # fix some vertices, connect shortest paths between them
      iList=[18,34,52,63,18]
      path=np.zeros(nEdges)
      for j in range(len(iList)-1):
          pathList=extractPath(pred,iList[j],iList[j+1])
          path+=translatePath(pathList,edgeData)

      fig=plt.figure(figsize=(6,6),facecolor="w")
      ax=fig.add_subplot(aspect=1.)

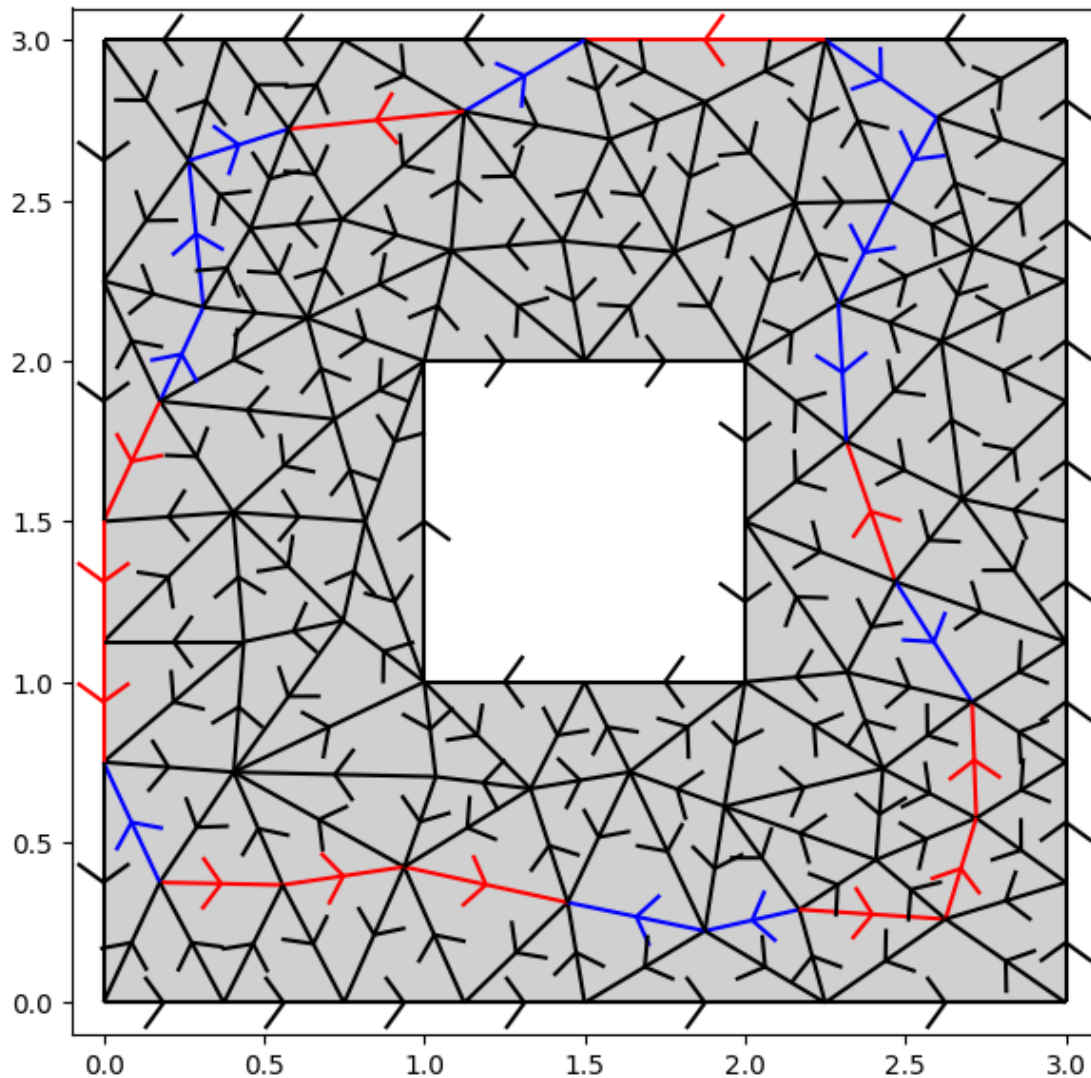
      polyCollection=matplotlib.collections.
          ↳LineCollection(pointData[triangleData],lw=0,fc="#d0d0d0",zorder=-2)
      ax.add_collection(polyCollection)
      lineColors=signToColor(path)
      drawEdges(ax,pointData,edgeData,colors=lineColors,drawOrientations=True,arrowScale=0.
          ↳1,zorder=1)

      #for i,x in enumerate(pointData):
      #    ax.text(x[0],x[1],i,c="r")

      # plot range
      buffer=0.1
      lim1,lim2,lim3,lim4=np.min(pointData[:,0]),np.max(pointData[:,0]),np.
          ↳min(pointData[:,1]),np.max(pointData[:,1])
      plt.xlim([lim1-buffer,lim2+buffer])
      plt.ylim([lim3-buffer,lim4+buffer])

      plt.tight_layout()
      plt.show()

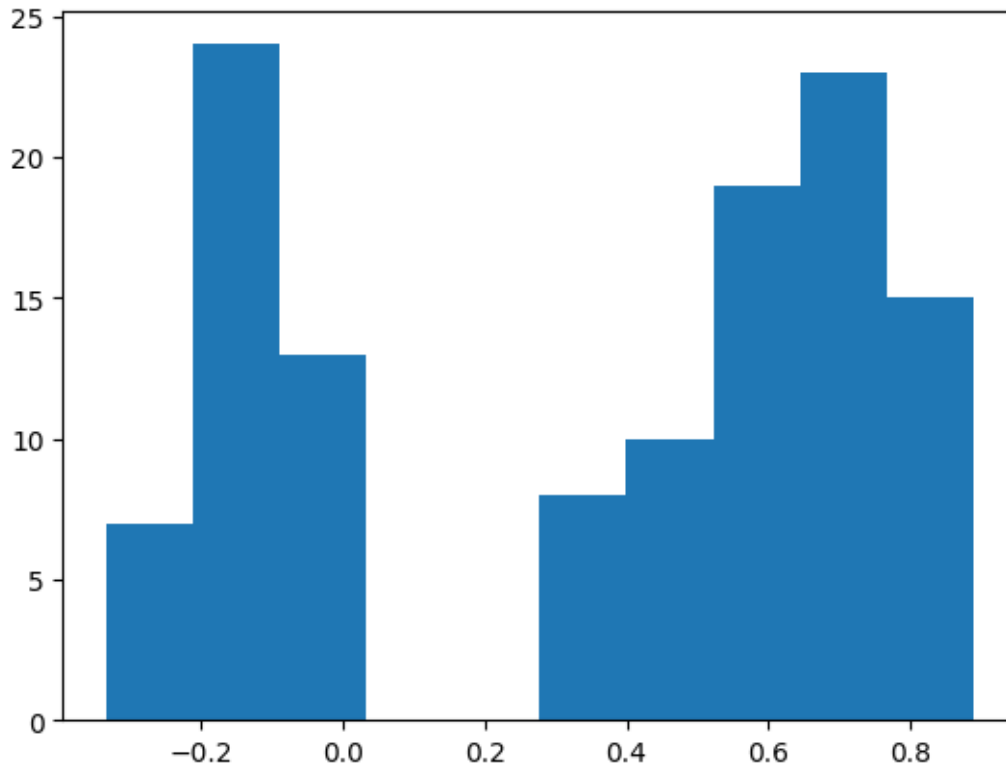
```



```
[40]: # as above: try to find region which has given path as boundary
# by trying to solve solve: B2.dot(region)=path
# since we do not know for sure if solution exists, use leastsquares ansatz
res=scipy.sparse.linalg.lsqr(B2,path)
region=res[0]
# test solution quality: this time we have a non-zero residual!
print("norm of discrepancy between B2(region) and path:",res[3])
print("if solution was found, this should be 1: ",res[1])
```

norm of discrepancy between B2(region) and path: 1.4968407965017811  
if solution was found, this should be 1: 2

```
[41]: # if a good solution was found, then region should be binary, with only 0 and 1
      ↪ entries
      plt.hist(region)
      plt.show()
```



```
[42]: # now add the boundary of the hole as additional column to B2:

# first: extract boundary of hole
iList=[4,5,7,6,4]
holeBoundary=np.zeros(nEdges)
for j in range(len(iList)-1):
    pathList=extractPath(pred,iList[j],iList[j+1])
    holeBoundary+=translatePath(pathList,edgeData)

# now add to B2:
B2Ext=scipy.sparse.hstack((B2,scipy.sparse.csr_matrix(holeBoundary.
    ↪ reshape((-1,1)))))
```

```
[43]: # do the rank test again:

# recall:
# -image of B2 lives in kernel of B1
```

```

# -compare dimensions:
#   dim(ker B1)=nEdges-rank(B1)
#   dim(img B2)=rank(B2)

# since we use small matrices: determine rank on dense matrices
B1Dense=B1.todense()
B2Dense=B2Ext.todense()

print(nEdges-np.linalg.matrix_rank(B1Dense))
print(np.linalg.matrix_rank(B2Dense))

```

120

120

```

[44]: # do least squares solving again:
res=scipy.sparse.linalg.lsqr(B2Ext,path)
region=res[0]
# test solution quality:
print("norm of discrepancy between B2(region) and path:",res[3])
print("if solution was found, this should be 1: ",res[1])

```

norm of discrepancy between B2(region) and path: 6.776446353336523e-05

if solution was found, this should be 1: 1

```

[45]: iList=[18,34,52,63,18]
path=np.zeros(nEdges)
for j in range(len(iList)-1):
    pathList=extractPath(pred,iList[j],iList[j+1])
    path+=translatePath(pathList,edgeData)

pathExt=path.copy()
pathExt+=holeBoundary*region[-1]

fig=plt.figure(figsize=(6,6),facecolor="w")
ax=fig.add_subplot(aspect=1.)

regionColors=np.zeros((nTriangles,3))
regionColors[...]=np.array([0.85,0.85,0.85])

#
regionColors[(region[:-1]>0.1),:]=np.array([0.5,0.5,0.5])
polyCollection=matplotlib.collections.
    ↳LineCollection(pointData[triangleData],lw=0,fc=regionColors,zorder=-2)
ax.add_collection(polyCollection)
lineColors=signToColor(pathExt)

```

```

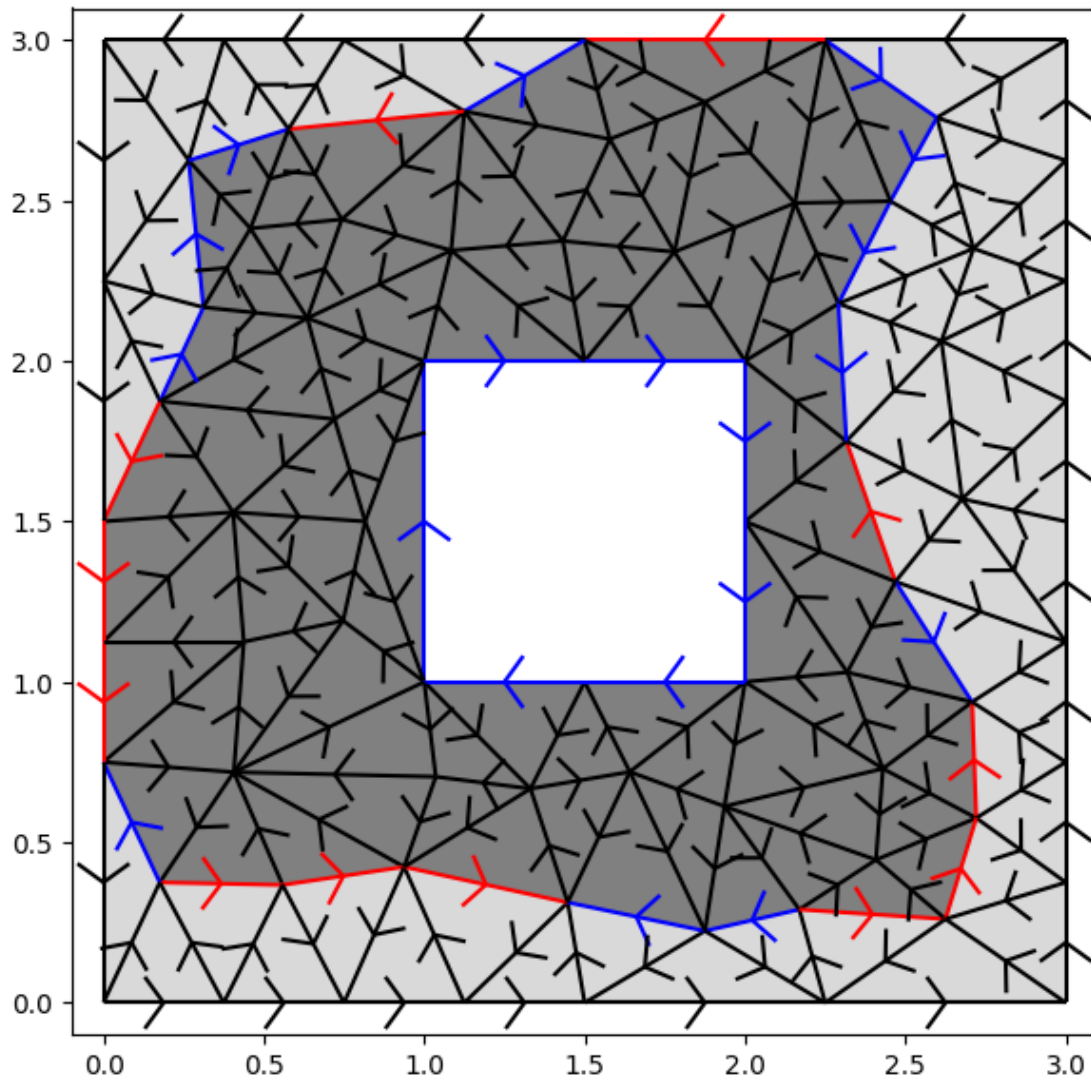
drawEdges(ax,pointData,edgeData,colors=lineColors,drawOrientations=True,arrowScale=0.
↪1,zorder=1)

#for i,x in enumerate(pointData):
#    ax.text(x[0],x[1],i,c="r")

# plot range
buffer=0.1
lim1,lim2,lim3,lim4=np.min(pointData[:,0]),np.max(pointData[:,0]),np.
↪min(pointData[:,1]),np.max(pointData[:,1])
plt.xlim([lim1-buffer,lim2+buffer])
plt.ylim([lim3-buffer,lim4+buffer])

plt.tight_layout()
plt.show()

```



1.10 Now repeat this on the 2-torus

```
[46]: def getPoslistNCube(shape, dtype=np.double):
    """Create list of positions in an n-dimensional cuboid of size shape."""
    ndim=len(shape)

    axGrids=[np.arange(i, dtype=dtype) for i in shape]
    prePos=np.array(np.meshgrid(*axGrids, indexing='ij'), dtype=dtype )
    # the first dimension of prepos is the dimension of the posvector, the
    ↪ successive dimensions are in the cube
    # so need to move first axis to end, and then flatten
    pos=np.rollaxis(prePos, 0, ndim+1)
```



```

# flattening
newshape=(-1,ndim)
return (pos.reshape(newshape)).copy()

```

```

[47]: # first build regular triangulation of rectangle
nRows=4
nCols=6

pointData=getPoslistNCube((nCols,nRows))

nTriangles=(nRows-1)*(nCols-1)*2
triangleData=np.zeros((nTriangles,3),dtype=np.int32)
k=0
for i in range(nCols-1):
    for j in range(nRows-1):
        ind0=i*nRows+j
        ind1=(i+1)*nRows+j
        ind2=(i+1)*nRows+(j+1)
        ind3=i*nRows+(j+1)
        triangleData[k]=np.array([ind0,ind1,ind2])
        k+=1
        triangleData[k]=np.array([ind0,ind2,ind3])
        k+=1

nPoints=pointData.shape[0]

# extract edge data
edgeData,etAdjacencyData,etAdjacencyDataOrientation=FEM.getEdges(triangleData)
nEdges=edgeData.shape[0]

```

```

[48]: %matplotlib inline
fig=plt.figure(figsize=(6,6),facecolor="w")
ax=fig.add_subplot(aspect=1.)

polyCollection=matplotlib.collections.
    ↳LineCollection(pointData[triangleData],lw=0,fc="#d0d0d0",zorder=-2)
ax.add_collection(polyCollection)
drawEdges(ax,pointData,edgeData,colors="k",drawOrientations=True,arrowScale=0.1)

for i,x in enumerate(pointData):
    ax.text(x[0],x[1],i,c="r",size=12)

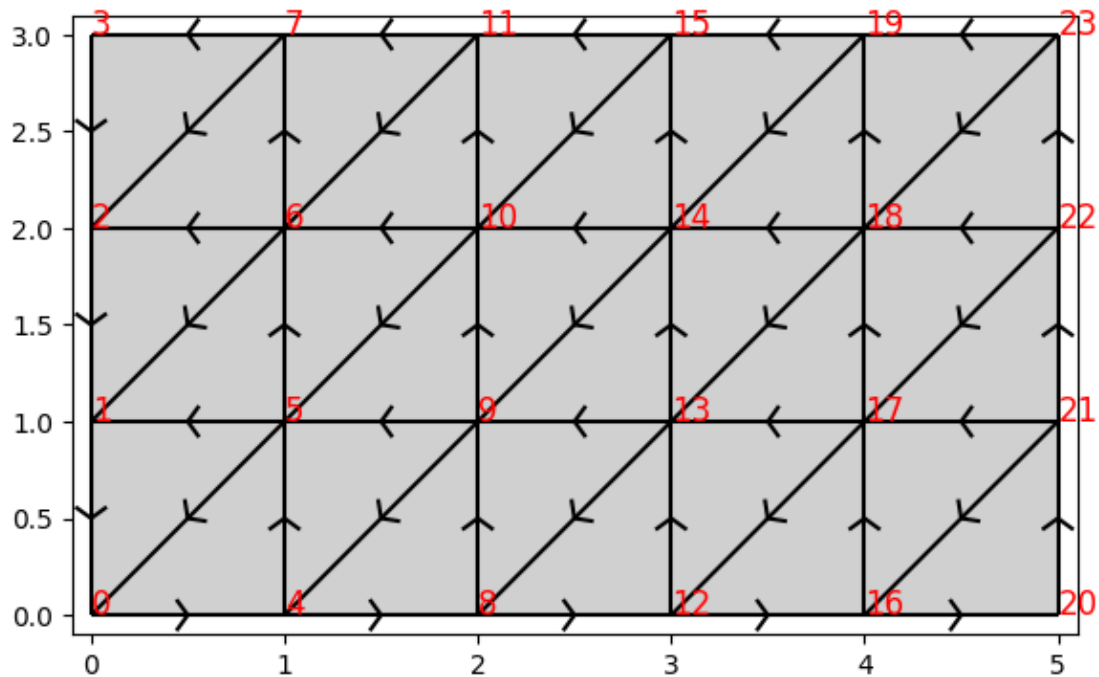
# plot range
buffer=0.1
lim1,lim2,lim3,lim4=np.min(pointData[:,0]),np.max(pointData[:,0]),np.
    ↳min(pointData[:,1]),np.max(pointData[:,1])
plt.xlim([lim1-buffer,lim2+buffer])

```

```
plt.ylim([lim3-buffer,lim4+buffer])
```

```
plt.tight_layout()
```

```
plt.show()
```



```
[49]: # Euler characteristic is that of planar mesh (not counting the "rest of
↪space"-face)
nTriangles-nEdges+nPoints
```

```
[49]: 1
```

```
[50]: # now build mesh for torus
nRows=10
nCols=20

nTriangles=nRows*nCols*2
triangleData=np.zeros((nTriangles,3),dtype=np.int32)
k=0
for i in range(nCols):
    for j in range(nRows):
        ind0=i*nRows+j
        ind1=((i+1)%nCols)*nRows+j
        ind2=((i+1)%nCols)*nRows+((j+1)%nRows)
```

```

        ind3=i*nRows+((j+1)%nRows)
        triangleData[k]=np.array([ind0,ind1,ind2])
        k+=1
        triangleData[k]=np.array([ind0,ind2,ind3])
        k+=1

nPoints=nRows*nCols

# extract edge data
edgeData,etAdjacencyData,etAdjacencyDataOrientation=FEM.getEdges(triangleData)
nEdges=edgeData.shape[0]

```

```

[51]: # Euler characteristic:
      nTriangles-nEdges+nPoints

```

```

[51]: 0

```

```

[52]: # at this point the mesh is only abstract, we have no coordinates
      # now build the well-known "donut embedding"
      pointData=np.zeros((nPoints,3),dtype=np.double)

      # big and small radii:
      R=1.
      r=0.3
      k=0
      for i in range(nCols):
          for j in range(nRows):
              # "large angle"
              phi=i/nCols*2*np.pi
              theta=j/nRows*2*np.pi

              n1=np.array([np.cos(phi),np.sin(phi),0.])
              n2=np.array([0.,0.,1.])
              pointData[k,:]=R*n1+r*np.cos(theta)*n2+r*np.sin(theta)*n1
              k+=1

```

```

[53]: import mpl_toolkits

```

```

[56]: %matplotlib inline
      fig = plt.figure(figsize=(8,8))

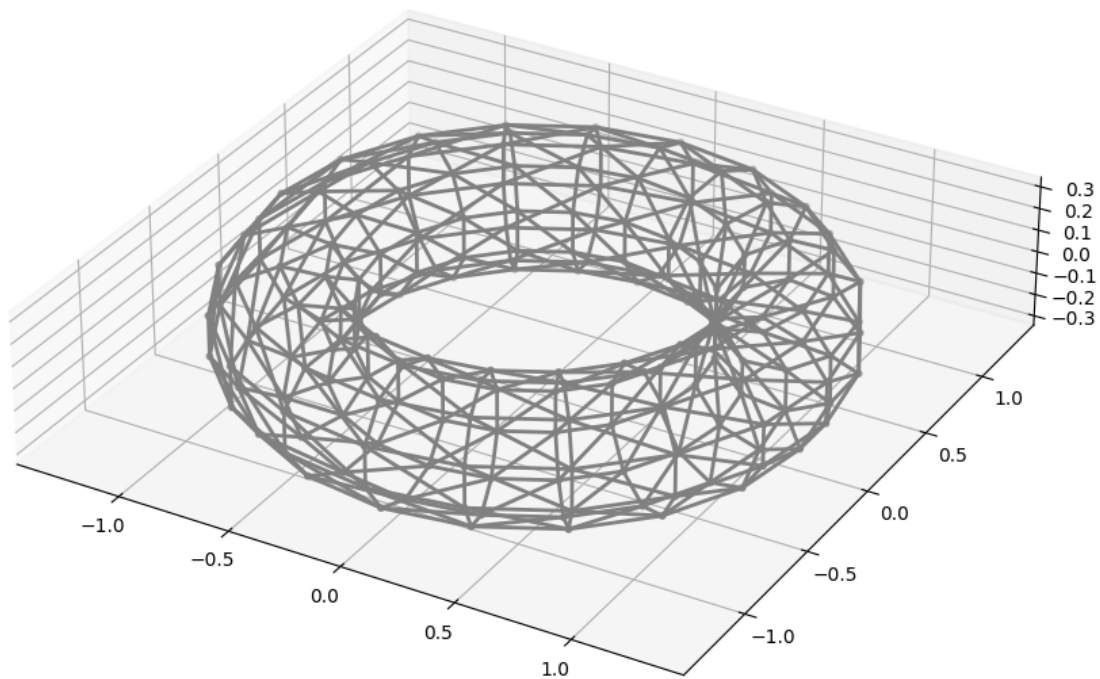
      ax = fig.add_subplot(projection='3d')

      ax.scatter(pointData[:,0],pointData[:,1],pointData[:,2],s=10,c="#808080")
      lineCollection=mpl_toolkits.mplot3d.art3d.
          ↪Line3DCollection(pointData[edgeData],color="#808080",lw=2)
      ax.add_collection(lineCollection)

```

```
ax.set_box_aspect((R+r,R+r,r))

plt.tight_layout()
plt.show()
```



```
[57]: B1=getB1(nPoints,edgeData)
      B2=getB2(nEdges,etAdjacencyData,etAdjacencyDataOrientation)
```

```
[58]: # do the rank test again:

      # recall:
      # -image of B2 lives in kernel of B1
```

```

# -compare dimensions:
#   dim(ker B1)=nEdges-rank(B1)
#   dim(img B2)=rank(B2)

# since we use small matrices: determine rank on dense matrices
B1Dense=B1.todense()
B2Dense=B2.todense()

print(nEdges-np.linalg.matrix_rank(B1Dense))
print(np.linalg.matrix_rank(B2Dense))

```

401  
399

```

[59]: path1Pre=np.arange(nRows+1)
      path1Pre[-1]=0
      path1=translatePath(path1Pre,edgeData)

      path2Pre=np.arange(nCols+1)*nRows
      path2Pre[-1]=0
      path2=translatePath(path2Pre,edgeData)

```

```

[62]: %matplotlib inline
      fig = plt.figure(figsize=(8,8))

      ax = fig.add_subplot(projection='3d')

      ax.scatter(pointData[:,0],pointData[:,1],pointData[:,2],s=10,c="#808080")

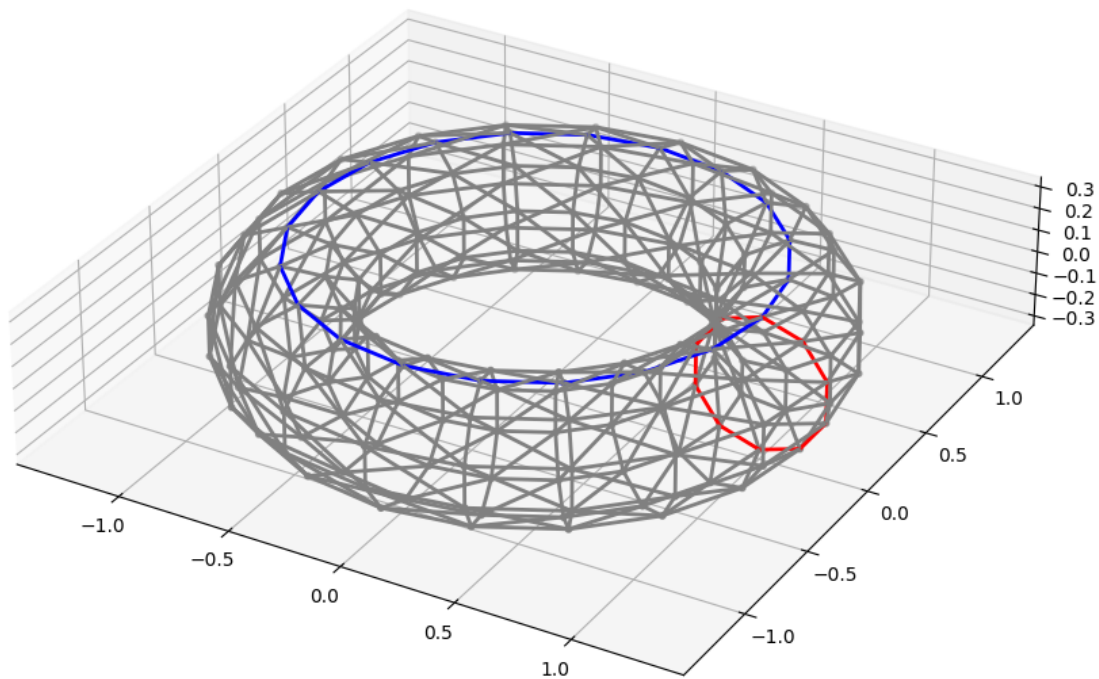
      lineColors=np.zeros((nEdges,3))
      lineColors[...] =np.array([0.5,0.5,0.5])
      lineColors[(np.abs(path1[:])>0.1),:]=np.array([1.,0.,0.])
      lineColors[(np.abs(path2[:])>0.1),:]=np.array([0.,0.,1.])

      lineCollection=mpl_toolkits.mplot3d.art3d.
        ↪Line3DCollection(pointData[edgeData],color=lineColors,lw=2)
      ax.add_collection(lineCollection)

      ax.set_box_aspect((R+r,R+r,r))

      plt.tight_layout()
      plt.show()

```



```
[63]: # add these as extra columns to B2:
B2Ext=scipy.sparse.hstack((B2,\
    scipy.sparse.csr_matrix(path1.reshape((-1,1))),\
    scipy.sparse.csr_matrix(path2.reshape((-1,1)))\
))
```

```
[64]: # do the rank test again:

# recall:
# -image of B2 lives in kernel of B1
# -compare dimensions:
#   dim(ker B1)=nEdges-rank(B1)
#   dim(img B2)=rank(B2)
```

```
# since we use small matrices: determine rank on dense matrices
B1Dense=B1.todense()
B2Dense=B2Ext.todense()

print(nEdges-np.linalg.matrix_rank(B1Dense))
print(np.linalg.matrix_rank(B2Dense))
```

401

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```
[65]: B1.dot(B2Ext)
```

```
[65]: <200x402 sparse matrix of type '<class 'numpy.float64'>'
      with 0 stored elements in Compressed Sparse Row format>
```

```
[ ]:
```