Documentation for graphcurvature.py

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Abstract

This documentation gives an overview of the file graphcurvature.py.

1 Introduction

Let G = (V, E) be a finite simple graph. For any vector (function) $f : V \to \mathbb{R}$ and any vertex $x \in V$, the Laplacian Δ is defined via

$$\Delta f(x) := \frac{1}{\mu(x)} \sum_{y, y \sim x} (f(y) - f(x)), \tag{1.1}$$

where $\mu : V \to \mathbb{R}$ is a positive measure on *V*. When $\mu(x) = 1$, for any $x \in V$, we call Δ the non-normalized Laplacian. When $\mu(x) = d_x := \sum_{y,y \sim x} 1$, for any $x \in V$, we call Δ the normalized Laplacian.

For any two functions $f, g: V \to \mathbb{R}$, we define two operators Γ and Γ_2 as follows:

$$2\Gamma(f,g) := \Delta(fg) - f\Delta g - (\Delta f)g, \tag{1.2}$$

$$2\Gamma_2(f,g) := \Delta(\Gamma(f,g)) - \Gamma(f,\Delta g) - \Gamma(\Delta f,g).$$
(1.3)

Note that $\Gamma(f, f)(x)$ and $\Gamma_2(f, f)(x)$ can be represented as matrices defined on the twoball of *x* and acting on *f* restricted to the two-ball. We denote these matrices as $\Gamma(x)$ and $\Gamma_2(x)$. **Definition 1.1.** Let $\mathcal{K} \in \mathbb{R}$ and $\mathcal{N} \in \mathbb{R}_+$. We say that the graph G = (V, E) satisfies the curvature-dimesion inequality (CD inequality) $CD(\mathcal{K}, \mathcal{N})$, if for any $f : V \to \mathbb{R}$ and any $x \in V$, we have

$$\Gamma_2(f)(x) \ge \frac{1}{\mathcal{N}} (\Delta f(x))^2 + \mathcal{K} \Gamma(f)(x).$$
(1.4)

Here, \mathcal{K} is called a lower Ricci curvature bound of G = (V, E), and \mathcal{N} a dimension parameter. At a vertex $x \in V$, the precise \mathcal{N} -dimensional Ricci curvature lower bound $\mathcal{K}_{\mathcal{N}}(G, x)$ is defined as the largest \mathcal{K} such that (1.4) holds for a given \mathcal{N} .

2 The Programme

2.1 Calculating the curvature

We enter graphs into Python via their adjacency matrix. For example, a triangle, i.e. the complete graph on 3 vertices, is entered as follows:

>>> T = [[0,1,1],[1,0,1],[1,1,0]]

Similarly the Petersen graph would be entered as:

>>> P=[[0,1,0,0,1,1,0,0,0,0],[1,0,1,0,0,0,1,0,0,0],[0,1,0,1,0,0,0,1,0,0], [0,0,1,0,1,0,0,0,1,0],[1,0,0,1,0,0,0,0,0,1],[1,0,0,0,0,0,0,0,1,1,0], [0,1,0,0,0,0,0,0,1,1],[0,0,1,0,0,1,0,0,0,1],[0,0,0,1,0,1,1,0,0,0], [0,0,0,0,1,0,1,1,0,0]]

The functions curv_calc and curv_calc_norm calculate the curvature at a specified vertex with respect to the non-normalised and normalised Laplacian, respectively. Note that the vertices are specified via their vertex numbers and that the enumeration starts from 0.

```
>>> curv_calc(T, 0)
2.5
>>> curv_calc_norm(T, 0)
1.25
>>> curv_calc(T, 0)
-1.0
>>> curv_calc_norm(T, 0)
-0.33
```

2.2 The matrices Γ and Γ_2

In this section we discuss the commands generating the matrices Γ and Γ_2 . The matrices are viewed conveniently via the numpy command array.

Generally 4Γ and $4\Gamma_2$ instead of Γ and Γ_2 ensures that all the terms in the matrices with respect to the non-normalised laplacian are integer valued. Again, the vertex in these commands is specified by its vertex number. The commands are fourGamma, fourGamma2, fourGammaNorm and fourGamma2Norm.

```
>>> np.array(fourGamma(T, 0))
[[ 4 -2 -2]
 [-2 2 0]
 [-2 0 2]]
>>> np.array(fourGamma2(T, 0))
[[10 -5 -5]
 [-5 7 -2]
 [-5 -2 7]]
>>> np.array(fourGammaNorm(T, 0))
[[ 2. -1. -1.]
 [-1. 1.
           0.]
 [-1. 0.
           1.]]
>>> np.array(fourGamma2Norm(T, 0))
[[ 2.5 -1.25 -1.25]
 [-1.25 1.75 -0.5]
 [-1.25 -0.5
               1.75]]
```

Note that the Γ - and Γ_2 -matrices have a different vertex ordering than the original adjacency matrix. It is convenient to build up these matrices by rearranging the vertices in the order "centre, one-sphere, two-sphere". The command mat_order reveals this order for a specified vertex in form of the original vertex numbers.

>>> np.array(fourGamma(P,								0))	
[[6	-2	-2	-2	0	0	0	0	0	0
[-2	2	0	0	0	0	0	0	0	0
[-2	0	2	0	0	0	0	0	0	0
[-2	0	0	2	0	0	0	0	0	0
[0	0	0	0	0	0	0	0	0	0
[0	0	0	0	0	0	0	0	0	0

```
[ 0
     0 0
           0
               0
                  0
                     0
                        0
                              0]
                           0
 [ 0 ]
     0
                     0
                              0]
        0
           0
               0
                 0
                        0
                           0
 [ 0 ]
     0
                     0
                              0]
         0
           0
               0
                 0
                        0
                           0
 [00]
           0
               0
                 0
                     0
                       0
                           0
                              0]]
        0
>>> np.array(fourGamma2(P, 0))
[[18 -8 -8 -8
              1
                     1
                  1
                        1
                           1
                              1]
           2 -2
 [-8 8
         2
                 0 -2
                        0
                           0
                              0]
 [-8 2
        8
           2 0 -2
                     0
                        0
                          0 -2]
 [-8 2
         2
           8
               0
                     0 -2 -2
                              0]
                 0
 [ 1 -2
        0
           0
               1
                 0
                     0
                        0
                           0
                              0]
 [1 0 -2
           0
               0
                  1
                     0
                        0
                           0
                              0]
 [ 1 -2
           0
               0
                  0
                     1
                        0
                           0
                              0]
       0
 [1 0 0 -2
               0
                  0
                     0 1
                              0]
                          0
 Γ1
     0 0 -2
              0
                 0
                     0
                        0
                          1
                              0]
 [ 1 0 -2 0 0 0
                     0
                        0
                          0
                             1]]
>>> mat_order(P, 0)
[0, [1, 4, 5], [2, 3, 6, 7, 8, 9]]
```

Note that you can also use this function to calculate the one-sphere and two-sphere of a given vertex.

Finally, the command evs calculates the eigenvalues of a given symmetric matrix. For example

>>> evs(fourGamma(T,0)) [0. 2. 6.]

3 Edge and Vertex weights

The functions fourGammaFULL(W, MU, i) and fourGamma2FULL(W, MU, i) calculate the matrices, defined in (1.2) and (1.3), with edge weight matrix W and a vector containing the vertex weights *MU*.

In this case the Laplacian assumes the form

$$\Delta f(x) := \frac{1}{\mu(x)} \sum_{y, y \sim x} w_{xy}(f(y) - f(x)), \tag{3.1}$$

with $w_{xy} = W(x, y)$ and $\mu(x) = MU(x)$.

For example, suppose we want to enter a triangle with edges weights 1,2 and 3 and calculate its $4\Gamma_2$ matrices about each vertex. We would do this as follows:

```
>>> T = [[0,1,2],[1,0,3],[2,3,0]]
>>> MU = [1, 1, 1]
>>> np.array(fourGammaFULL(T, MU, 0))
[[ 6. -2. -4.]
 [-2.
      2.
           0.]
 [-4.
       0.
           4.]]
>>> np.array(fourGammaFULL(T, MU, 1))
[[ 8. -2. -6.]
 [-2.
      2.
           0.]
 [-6.
           6.]]
       0.
>>> np.array(fourGammaFULL(T, MU, 2))
[[ 10. -4.
             -6.]
 [ -4.
         4.
              0.]
 Γ-6.
              6.]]
         0.
```

Note that if you choose the adjacency matrix for *W* then you obtain the Laplacian defined in equation (1.1).

4 Interactive website

In addition to the python programme there is an interactive website which allows you to draw a graph and to view the curvature of the graph at each vertex with respect to both the normalised and non-normalised Laplacian. Online instructions can be found on the webpage itself.

To play with it go to http://teggers.eu/graph/.