

7. Matrix examples

7.3. Incidence matrix

Incidence matrix of a graph. We create the incidence matrix of the network shown in Figure 7.3 in VMLS.

```
In [ ]: A = np.array([[ -1, -1, 0, 1, 0], [ 1, 0, -1, 0, 0], [ 0, 0, 1, -1, -1],  
    ↪ [ 0, 1, 0, 0, 1]])  
xcirc = np.array([1, -1, 1, 0, 1]) #A circulation  
A @ xcirc
```

```
Out [ ]: array([0, 0, 0, 0])
```

```
In [ ]: s = np.array([1, 0, -1, 0, 1]) # A source vector  
x = np.array([0.6, 0.3, 0.6, -0.1, -0.3]) #A flow vector  
A @ x + s #Total incoming flow at each node
```

```
Out [ ]: array([1.11022302e-16, 0.00000000e+00, 0.00000000e+00,  
    ↪ 0.00000000e+00])
```

Dirichlet energy. On page 135 of VMLS we compute the Dirichlet energy of two potential vectors associated with the graph of Figure 7.2 in VMLS.

```
In [ ]: A = np.array([[ -1, -1, 0, 1, 0], [ 1, 0, -1, 0, 0], [ 0, 0, 1, -1, -1],  
    ↪ [ 0, 1, 0, 0, 1]])  
vsmooth = np.array([1, 2, 2, 1])  
np.linalg.norm(A.T @ vsmooth)**2 #Dirichlet energy of vsmooth
```

```
Out [ ]: 2.9999999999999996
```

```
In [ ]: vrough = np.array([1, -1, 2, -1])  
np.linalg.norm(A.T @ vrough)**2 # Dirichlet energy of vrough
```

```
Out [ ]: 27.0
```

7.4. Convolution

The numpy function `np.convolve()` can be used to compute the convolution of the vectors `a` and `b`. Let's use this to find the coefficients of the polynomial

$$p(x) = (1+x)(2-x+x^2)(1+x-2x^2) = 2 + 3x - 3x^2 - x^3 + x^4 - 2x^5$$

```
In [ ]: a = np.array([1,1]) # coefficients of 1+x
        b = np.array([2,-1,1]) # coefficients of 2-x+x^2
        c = np.array([1,1,-2]) # coefficients of 1+x-2x^2
        d = np.convolve(np.convolve(a,b),c) # coefficients of product
        d

Out [ ]: array([ 2,  3, -3, -1,  1, -2])
```

Let's write a function that creates a Toeplitz matrix and check it against the `conv` function. We will also confirm that Python is using a very efficient method for computing the convolution.

To construct the Toeplitz matrix $T(b)$ defined in equation (7.3) of VMLS, we can first create a zero matrix of the correct dimensions $((n + m - 1) \times n)$ and then add the coefficients b_i one by one. Single-index indexing comes in handy for this purpose. The single-index indexes of the elements b_i in the matrix $T(b)$ are $i, i + m + n, i + 2(m + n), \dots, i + (n - 1)(m + n)$.

```
In [ ]: b = np.array([-1,2,3])
        a = np.array([-2,3,-1,1])
        def toeplitz(b,n):
            m = len(b)
            T = np.zeros((n+m-1,n))
            for j in range(n):
                T[j:j+m,j] = b
            return T
        Tb = toeplitz(b,len(a))
        Tb
```

```
Out [ ]: array([[ -1.,  0.,  0.,  0.],
               [  2., -1.,  0.,  0.],
               [  3.,  2., -1.,  0.],
               [  0.,  3.,  2., -1.],
               [  0.,  0.,  3.,  2.],
               [  0.,  0.,  0.,  3.]])
```

```
In [ ]: Tb @ a, np.convolve(b,a)
```

```
Out [ ]: (array([2.,-7., 1., 6.,-1., 3.]), array([2,-7, 1, 6,-1, 3]))
```

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```
In [ ]: import time
        m = 2000
        n = 2000
        b = np.random.normal(size = n)
        a = np.random.normal(size = m)
        start = time.time()
        ctoep = toeplitz(b,n) @ a
        end = time.time()
        print(end - start)
```

```
0.05798077583312988
```

```
In [ ]: start = time.time()
        cconv = np.convolve(a,b)
        end = time.time()
        print(end - start)
```

```
0.0011739730834960938
```

```
In [ ]: np.linalg.norm(ctoep - cconv)
```

```
Out [ ]: 1.0371560230890881e-12
```