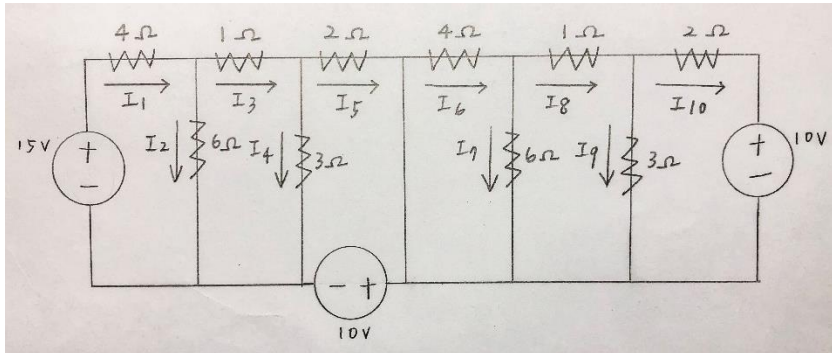


Linear Algebra Assignment 2 Report

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(a) your circuit



$$I_1 = I_2 + I_3$$

$$I_6 = I_7 + I_8$$

$$I_3 = I_4 + I_5$$

$$I_8 = I_9 + I_{10}$$

$$4I_1 + 6I_2 = 15$$

$$4I_6 + 6I_7 = 10$$

$$6I_2 - I_3 - 3I_4 = 0$$

$$6I_7 - I_8 - 3I_9 = 0$$

$$3I_4 - 2I_5 = 10$$

$$3I_9 - 2I_{10} = 10$$

(b) your matrix A and b

$$A = \begin{bmatrix} 1 & -1 & -1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & -1 & -1 & 0 & 0 & 0 & 0 & 0 \\ 4 & 6 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 6 & -1 & -3 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 3 & -2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & -1 & -1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & -1 & -1 \\ 0 & 0 & 0 & 0 & 0 & 4 & 6 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 6 & -1 & -3 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 3 & -2 \end{bmatrix}$$

$$B = [[0.0],[0.0],[15.0],[0.0],[10.0],[0.0],[0.0],[0.0],[0.0],[10.0]]$$

(c) $A^{-1} =$

$$\begin{bmatrix} 0.29 & 0.16 & 0.18 & -0.13 & -0.08 & -0. & 0. & -0. & 0. & -0. \\ -0.19 & -0.1 & 0.05 & 0.09 & 0.05 & 0. & -0. & 0. & -0. & 0. \\ -0.52 & 0.26 & 0.13 & -0.22 & -0.13 & -0. & 0. & -0. & 0. & -0. \\ -0.21 & -0.3 & 0.05 & -0.09 & 0.15 & 0. & -0. & 0. & -0. & 0. \\ -0.31 & -0.44 & 0.08 & -0.13 & -0.28 & -0. & 0. & -0. & 0. & -0. \\ -0. & 0. & -0. & 0. & -0. & 0.29 & 0.16 & 0.18 & -0.13 & -0.08 \\ -0. & -0. & 0. & -0. & 0. & -0.19 & -0.1 & 0.05 & 0.09 & 0.05 \\ -0. & 0. & -0. & 0. & -0. & -0.52 & 0.26 & 0.13 & -0.22 & -0.13 \\ -0. & -0. & 0. & -0. & 0. & -0.21 & -0.3 & 0.05 & -0.09 & 0.15 \\ -0. & 0. & -0. & 0. & -0. & -0.31 & -0.44 & 0.08 & -0.13 & -0.28 \end{bmatrix}$$

```

[[ 1.89]
 [ 1.24]
 [ 0.65]
 [ 2.26]
 [-1.61]
 [-0.78]
 [ 0.52]
 [-1.3 ]
 [ 1.48]
X= [-2.78]]

```

(d) execution results for the third problem.(TA judge)

```

(2 x 2)
(mydetA): Success
(mysolve_cramer): Success
[[ 1.1 ]
 [-0.68]]
(mysolve_adj): Success
(3 x 3)
(mydetA): Success
(mysolve_cramer): Success
[[ -0.96]
 [-1.97]
 [ 4.21]]
(mysolve_adj): Success
(4 x 4)
(mydetA): Success
(mysolve_cramer): Success
[[ 10.83]
 [ 20.97]
 [ 4.9 ]
 [-18.79]]
(mysolve_adj): Success
(5 x 5)
(mydetA): Success
(mysolve_cramer): Success
[[ 2.28]
 [ 0.55]
 [-2.2 ]
 [ 0.43]
 [ 0.4 ]]
(mysolve_adj): Success
(6 x 6)
(mydetA): Success
(mysolve_cramer): Success
[[ 1.66]
 [-2.03]
 [-0.47]
 [ 1.27]
 [ 0.3 ]
 [ 0.29]]
(mysolve_adj): Success
(7 x 7)
(mydetA): Success
(mysolve_cramer): Success
[[ -0.07]
 [-0.17]
 [ 0.28]
 [ 0.57]
 [-0.45]
 [ 0.78]
 [ 0.35]]
(mysolve_adj): Success

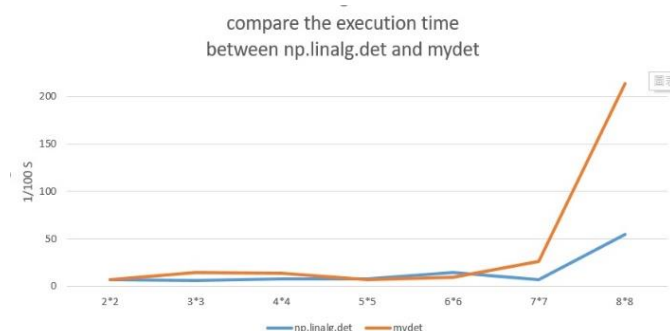
```

(Judge Test) Total Score: 18/18

(e) 比較 np.linalg.det 和 mydet 的執行時間

從2*2的矩陣到10*10的矩陣，可發現使用 np.linalg.det 執行時間大多比 mydet 快，且矩陣越大執行時間差越大。

```
(2 x 2)
0.2678103598726722
Using np.linalg.det,time to solve det is 0:00:00.007016 seconds.
0.2678103598726722
Using mydet,time to solve det is 0:00:00.007137 seconds.
(3 x 3)
0.027352149988940814
Using np.linalg.det,time to solve det is 0:00:00.005980 seconds.
0.027352149988940828
Using mydet,time to solve det is 0:00:00.014504 seconds.
(4 x 4)
0.030217798576239057
Using np.linalg.det,time to solve det is 0:00:00.007469 seconds.
0.03021779857623906
Using mydet,time to solve det is 0:00:00.013463 seconds.
(5 x 5)
-0.02553496564181468
Using np.linalg.det,time to solve det is 0:00:00.007980 seconds.
-0.025534965641814653
Using mydet,time to solve det is 0:00:00.007018 seconds.
(6 x 6)
0.002989638593517613
Using np.linalg.det,time to solve det is 0:00:00.014028 seconds.
0.002989638593517622
Using mydet,time to solve det is 0:00:00.009606 seconds.
(7 x 7)
0.006265436427282263
Using np.linalg.det,time to solve det is 0:00:00.006931 seconds.
0.006265436427282271
Using mydet,time to solve det is 0:00:00.026277 seconds.
(8 x 8)
-0.03521925762178388
Using np.linalg.det,time to solve det is 0:00:00.054221 seconds.
-0.0352192576217839
Using mydet,time to solve det is 0:00:00.213624 seconds.
(9 x 9)
0.0042511263867378935
Using np.linalg.det,time to solve det is 0:00:00.006038 seconds.
0.004251126386737917
Using mydet,time to solve det is 0:00:01.328202 seconds.
(10 x 10)
-0.004523545109875117
Using np.linalg.det,time to solve det is 0:00:00.005984 seconds.
-0.004523545109875182
Using mydet,time to solve det is 0:00:13.155335 seconds.
```



用我自己設的10*10矩陣，mydet較快。

```
52899.99999999993
Using numpy,time to solve det is 0:00:00.020343 seconds.
52900.0
Using mydet,time to solve det is 0:00:00.009016 seconds.
```

(f) 比較 np.linalg.solve、mysolve_adj、mysolve_cramer 的執行時間

因為有些function運算的執行時間太短，會顯示0，所以我運用迴圈(for i in range(0,100))將所有運算都跑100次，以下實際執行時間都要除以100才是正確的。很明顯的可以看出來，np.linalg.solve的執行速度小於其他兩者。

```
Using numpy,time to solve det is 0:00:00.000997 seconds.
Using mydet,time to solve det is 0:00:00.112700 seconds.
Using np.linalg.solve, time to solve Ax=b is 0:00:00.000883 seconds.
Using mysolve_adj, time to solve Ax=b is 0:00:04.832379 seconds.
Cramer's rule, Execution Time = 0:00:03.825193 seconds.
```

(g) residuals

運用 $\|Ax-b\|$ (歐式距離)，計算誤差，mysolve_adj 與 mysolve_cramer 皆和 np.linalg.solve 的誤差差不多，皆很小，因此可推論準確度皆很高。

```
Using numpy.time to solve det is 0:00:00 seconds.
Using mydet.time to solve det is 0:00:00.001031 seconds.
Using np.linalg.solve, time to solve Ax=b is 0:00:00 seconds.
residuals of np.linalg.solve
[[-2.22044605e-16]
 [ 4.44089210e-16]
 [ 0.00000000e+00]
 [-1.11022302e-16]
 [ 0.00000000e+00]
 [ 3.33066907e-16]
 [-4.44089210e-16]
 [ 0.00000000e+00]
 [ 0.00000000e+00]
 [ 0.00000000e+00]]
Using mysolve_adj, time to solve Ax=b is 0:00:00.051857 seconds.
residuals of mysolve_adj
[[-2.22044605e-16]
 [-4.44089210e-16]
 [ 0.00000000e+00]
 [ 2.22044605e-16]
 [ 0.00000000e+00]
 [ 0.00000000e+00]
 [-2.22044605e-16]
 [-4.44089210e-16]
 [ 0.00000000e+00]
 [ 0.00000000e+00]]
Execution Time = 0:00:00.038945 seconds.
residuals of mysolve_cramer
[[ 0.00000000e+00]
 [ 4.44089210e-16]
 [ 0.00000000e+00]
 [ 2.22044605e-16]
 [ 0.00000000e+00]
 [ 0.00000000e+00]
 [ 0.00000000e+00]
 [-4.44089210e-16]
 [ 0.00000000e+00]
 [ 0.00000000e+00]]
```

(h) Bonus

`numpy.linalg.det`是運用 **LU decomposition**這種演算法來解，LU decomposition就是先算上三角矩陣和下三角矩陣，然後用 $\det(A)=\det(P)\det(L)\det(U)$ 或 $\det(A)=\det(L)\det(U)$ (如果沒有permutation matrix)，在矩陣較大時可以節省很多時間。下圖為numpy函式庫內對於det的實作，註解中有寫到是使用LUdecomposition來做。

```
1970 def det(a):
1971     """
1972     Compute the determinant of an array.
1973
1974     Parameters
1975     -----
1976     a : (..., M, M) array_like
1977         Input array to compute determinants for.
1978
1979     Returns
1980     -----
1981     det : (...) array_like
1982         Determinant of 'a'.
1983
1984     See Also
1985     -----
1986     slogdet : Another way to represent the determinant, more suitable
1987         for large matrices where underflow/overflow may occur.
1988
1989     Notes
1990     -----
1991
1992     .. versionadded:: 1.8.0
1993
1994     Broadcasting rules apply, see the 'numpy.linalg' documentation for
1995     details.
1996
1997     The determinant is computed via LU factorisation using the LAPACK
1998     routine s/dgetrf.
1999
2000     Examples
2001     -----
2002     The determinant of a 2-D array  $\begin{bmatrix} a & b \\ c & d \end{bmatrix}$  is  $ad - bc$ :
2003
2004     >>> a = np.array([[1, 2], [3, 4]])
2005     >>> np.linalg.det(a)
2006     -2.0
2007
2008     Computing determinants for a stack of matrices:
2009
2010     >>> a = np.array([ [1, 2], [3, 4]], [ [1, 2], [2, 1]], [ [1, 3], [3, 1]] ])
2011     >>> a.shape
2012     (3, 2, 2)
2013     >>> np.linalg.det(a)
2014     array([-2., -3., -8.])
2015
2016     """
2017     a = asarray(a)
2018     _assertRankAtLeast2(a)
2019     _assertNdSquareness(a)
2020     t, result_t = _commonType(a)
2021     signature = 'D->D' if isComplexType(t) else 'd->d'
2022     r = _umath_linalg.det(a, signature=signature)
2023     r = r.astype(result_t, copy=False)
2024     return r
```

以下為一個LU decomposition的實例：

$$A = P.L.U$$

where

$$A = \begin{pmatrix} 4 & 3 & 2 & 1 \\ 1 & 10 & 3 & 4 \\ 5 & 3 & 2 & -4 \\ 4 & 8 & 7 & 9 \end{pmatrix}$$

$$P = \begin{pmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

$$L = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 4 & 1 & 0 & 0 \\ 4 & \frac{37}{32} & 1 & 0 \\ 5 & \frac{47}{32} & \frac{181}{135} & 1 \end{pmatrix}$$

$$U = \begin{pmatrix} 1 & 10 & 3 & 4 \\ 0 & -32 & -5 & -7 \\ 0 & 0 & -\frac{135}{32} & -\frac{221}{32} \\ 0 & 0 & 0 & -\frac{602}{135} \end{pmatrix}$$

(g) bonus2

在mydet中我有運用條件判斷if A[0][i]!=0，可以節省執行時間，因為如果有很多0(sparse matrix)，計算det+=A[0][i]*((-1)**i)*mydet(minor)時，會做很多不必要的運算，因為A[0][i]如果是0，整項都會變0，就不用再算該項的minor，節省不必要的時間浪費。

(h) bonus reference

Finding determinant of 4*4 Matrix via LU Decomposition?

<https://math.stackexchange.com/questions/831823/finding-determinant-of-44-matrix-via-lu-decomposition?fbclid=IwAR0RLFsgCs1nlffVO1ixlSa8PAJgtS1TM5Uj5S23e5e0joLAyy-cO6hP2-E>