Homework 1 answers

Note: For full credit, show your work!

You are welcome to discuss the problems with others, but write up your own solutions.

Tor the matrix

$$A = \begin{pmatrix} 100 & 1/3 & 1/3 & 1/3 \\ 0 & 10 & 1/3 & 1/3 & 1/3 \\ 0 & 0 & 1 & 1/3 & 1/3 & 1/3 \\ 0 & 0 & 0 & 1/3 & 1/3$$

First do the calculation in Matlab. Then do the same calculation by hand, justifying why the answer is correct.

	9	, ,		\mathcal{I}			
٨				PHA	\wedge		
Av	swer:		import nump	import numpy as np			
>> format ra	t				k([[np.eye(3),	np.ones((3,3))/3],	
>> n = 3; A = [eye(n) ones(n,n)/n ; zeros(n,n) ones(n,n)/n]				-)), np.ones((3,3))/3]])	
				np.linalg.m	np.linalg.matrix_power(A, 300)		
A =							
1	0	0	1/3	1/3	1/3		
0	1	0	1/3	1/3	1/3		
0	0	1	1/3	1/3	1/3		
0	0	0	1/3	1/3	1/3		
0 0	0 0	0 0	1/3 1/3	1/3 1/3	1/3		
Ø	v	Ø	1/3	1/3	1/3		
>> A^300							
ans =							
/ 1	a	а	100	100	100		
/ o	1	ø	100	100	100		
0 0 0 0	1 0 0	1	100	100	100		
0	0	0	1/3	1/3	1/3		
\ 0	0	0	1/3	1/3	1/3 1/3		
\ 0	v	О	1/3	1/3			
(L	mu? A is	a block	matix:				
	3		2 1000		-1 1		
1 (I B) where I is the 3×3 identity matrix							
Why? A is a block matrix: $A = \begin{pmatrix} I & B \\ O & B \end{pmatrix} \text{where } I \text{ is the } 3\times3 \text{ identity matrix}$ $B = \frac{1}{3} \begin{pmatrix} & & \\ & & & \\ & & & \\ & & $							
$B = \frac{1}{3}(111)$							
	,		. \	3 (-(11).		
$\Rightarrow 12 = (I I B + B^2)$ $\Rightarrow 12 = 12 = 12 = 12 = 12$							
$\Rightarrow A^{2} = \begin{pmatrix} I & I \cdot B + B^{2} \\ O & B^{2} \end{pmatrix}$ Since $B^{2} = B$, $A^{2} = \begin{pmatrix} I & 2B \\ O & B \end{pmatrix}$.							
	\0	13					
٨	Vow let us p Claim: Fo	nave la	roduction				
<i>)</i>	1000 ict dis	hiore ad	171000001 1011	, L / I	LB \		
	Claim: Fo	r k=1,	2,3,	A = 1	R).		
		,	, ,				
	Proof:						
	Base co	ise k=1:	. /	/-	- (L-I)R)		
$\frac{1}{1-\sqrt{1-\sqrt{1-\sqrt{1-\sqrt{1-\sqrt{1-\sqrt{1-\sqrt{1-\sqrt{1-\sqrt{1$							
Sase case $k=1:$ Induction step: Assume $A^{k-1} = \begin{pmatrix} I & (k-1)B \\ O & B \end{pmatrix}$.							
\(I (k-1)B\							

$$\Rightarrow A^{k} = A \cdot A^{k-1} = \begin{pmatrix} I & B \\ O & B \end{pmatrix} \begin{pmatrix} I & (k-1)B \\ O & B \end{pmatrix}$$

$$= \begin{pmatrix} I & (k-1)B + B^{2} \\ O & B^{2} \end{pmatrix} = \begin{pmatrix} I & kB \\ O & B \end{pmatrix} / \Box$$

② a) Give a
$$2\times2$$
 matrix A that transforms $(1,0)$ to (a,c)

Assuming ad ≠ bc, give a 2×2 matrix that transforms

(Feel free to use a computer, if that helps.)

c) Give a 3×3 matrix B that transforms

$$(2,-1,0)$$
 to $(15,0,0)$

(Hint: Use part (D!)

a)
$$A=\begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

$$A^{-1} = \frac{1}{ad-bc} \begin{pmatrix} d - b \\ -c a \end{pmatrix}$$

b)
$$A^{-1} = \begin{pmatrix} 1 & 2 & 0 \\ 1 & -1 & 3 \\ 1 & 0 & 3 \end{pmatrix} \implies A = \begin{pmatrix} 1 & 2 & -2 \\ 0 & -1 & 1 \\ -\frac{1}{3} & -\frac{1}{3} & 1 \end{pmatrix}$$

c)
$$\begin{pmatrix} 5 & 15 & -1 \\ 10 & 0 & -1 \\ -1 & 0 & 1 \end{pmatrix} \times \begin{pmatrix} 1 & 2 & -2 \\ 0 & -1 & 1 \\ -\frac{7}{3} & -\frac{7}{3} & 1 \end{pmatrix} = \begin{pmatrix} 16/3 & -13/3 & 4 \\ 31/3 & 62/3 & -21 \\ -4/3 & -8/3 & 3 \end{pmatrix}$$

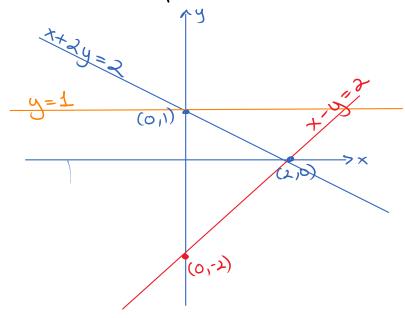
3 Sketch the three lines

$$x + \lambda y = \lambda$$

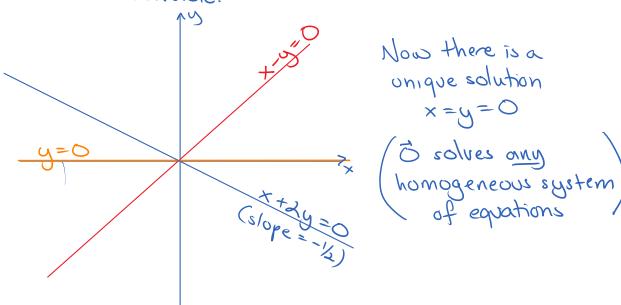
$$x - y = \lambda$$

$$y = 1$$

x-y=2 y=1 Are the equations solvable? What happens if the right-hand sides are all 0? Is there any nonzero choice for the gight-hand ordes that allows the three lines to intersect at the same point?



Since the three lines don't intersect, the equations are not solvable.



$$u+w+z=2$$
, and

u+w=2, all in four-dimensional space. Is it a line or a point or an empty set?

Answer: Use Gaussian elimination to find their intersection:

> the solution is

$$y=4$$
, $z=0$
 $u=2-\omega$, where ω is free
This is α line

What is the intersection if the fourth plane u=-1 is included? Find a fourth equation that

leaves us with no solution.

If u=-1 is added, the solution is a point: (u,v,w,z)=(-1,4,3,0)

u+w=10 will give no solution, for example

(5) How fast is Matlab?

A useful skill is to be able to estimate how long a calculation is going to take before starting it. (For

example, you can determine how large a problem you can solve without buying a new computer!)

a) Matrix multiplication:

Here is some crude Matlab/Octave code for timing the multiplication of two random 1000×1000 matrices:

```
% time to multiply two random 1000x1000 matrices
                                                      Puthon
n = 1000;
A = randn(n, n);
B = randn(n, n);
                                              import time
starttime = cputime;
                                              import numpy as np
                              CPU time
C = A * B;
                            in seconds
                                              n = 2000
endtime = cputime;
                                              trials = 10
elapsedtime = endtime - starttime
                                              totaltime = 0
% same code, but averaged over 10 trials
                                              for _ in range(trials):
                         (to reduce woise) A = np.random.randn(n, n)
trials = 10;
                                                B = np.random.randn(n, n)
n = 1000;
                                                start = time.perf_counter()
                                                C = A.dot(B) # or np.dot(A,B)
A = randn(n, n);
                                                              # NOT A * B!
B = randn(n, n);
                                                totaltime += time.perf_counter() - start
starttime = cputime;
                                              averagetime = totaltime / trials
for i = 1:trials
                                              averagetime
 C = A * B;
end
endtime = cputime;
```

averageelapsedtime = (endtime - starttime)/trials

Your problem: Estimate the scaling of the running time for watrix multiplication, as n increases. That is, if the running time is $\approx c \cdot n^{\alpha}$ for some exponent a and constant c, estimate values for a and c. (Don't worry about being too precise, but try to get something reasonable. For example, if $\alpha = 3$, then running the above code with n=2000 should take $\theta = 2^{\alpha}$ times as long. Show your work!)

Based on your estimates for c and x, for how large an n can your computer multiply two random n×n matrices in one day (24 hours)?

(Again, don't worry about being too precise, or about what

nappens when your computer runs out of memory.)

Answer: There are lots of ways of doing this, to find and a to fit observed running times. Well talk more about fitting curves later.

One simple solution to find x is to simply take the ratio of the average CPU time for n=1000 and n=2000, ie.,

$$\frac{t_{2000}}{t_{1000}} = \frac{\cancel{(2000)}^d}{\cancel{(1000)}^d} = 2^d$$

$$\Rightarrow 2 = \log_2(\frac{t_{2000}}{t_{1000}})$$

$$c = \frac{t_{1000}}{1000^d}$$

Since 24 hours = 86,400 seconds, solving

$$86400 = c \cdot n^{4}$$

$$\Rightarrow n = \left(\frac{86400}{c}\right)^{1/4}$$

My Matlab code:

```
n = 1000:
                                                    n = 2000;
totaltime = 0;
                                                    totaltime = 0:
                                                    trials = 10;
trials = 10;
for i = 1:trials
                                                    for i = 1:trials
 A = randn(n, n);
                                                      A = randn(n, n);
                                                      B = randn(n, n);
 B = randn(n, n);
 starttime = cputime;
                                                      starttime = cputime;
                                                      C = A * B;
 C = A * B;
                                                      endtime = cputime;
  endtime = cputime;
                                                      totaltime = totaltime + (endtime - starttime);
  totaltime = totaltime + (endtime - starttime);
                                                    t2000 = totaltime / trials
t1000 = totaltime / trials
                                                         alpha =
                      alpha = log2(t2000/t1000)
                                                            2.9069
                      c = t1000 / 1000^alpha
                      n24 = (86400 / c)^{(1/alpha)}
                                                            2.3211e-10
         Your answers will vary!
                                                            1.0292e+05 \approx 100,000
```

6) Solvina a sustain of linear equations:

6) Solving a system of linear equations:
Repeat part a, but for solving n random linear equations in n variables.

```
Answer: Just as above, heres my code:
                                                   n = 2000:
n = 1000;
                                                   totaltime = 0;
totaltime = 0;
                                                   trials = 10;
trials = 10;
                                                   for i = 1:trials
for i = 1:trials
                                                     A = randn(n, n);
  A = randn(n, n);
                                                     b = randn(n, 1);
  b = randn(n, 1);
                                                     starttime = cputime;
  starttime = cputime;
                                                     x = A \setminus b;
  x = A \setminus b;
                                                     engtime = cputime;
  endtime = cputime;
                                                     totaltime = totaltime + (endtime - starttime);
  totaltime = totaltime + (endtime - starttime);
                                                   t2000 = totaltime / trials
t1000 = totaltime / trials
                                                    ≨alpha =
                   alpha = log2(t2000/t1000)
                                                        2.4829
                   c = t1000 / 1000^alpha
                   n24 = (86400 / c)^{(1/alpha)}
                                                     c =
                                                         2.9529e-09
                                                      n24 =
                                                         2.6516e+05 ≈ ¾65,000
```

c) Solving a sparse system of linear equations: Repeat part a, but for solving

 $A\vec{x} = \vec{b}$,

where b is a random vector of length n, and A is an n*n matrix with 10n random nonzero entries in random positions.

This code might be helpful:

```
n = 1000;
                                                    n = 2000;
numnonzeroentries = 10 * n;
                                                    numnonzeroentries = 10 * n;
totaltime = 0;
                                                     totaltime = 0;
trials = 30;
                                                     trials = 30;
for i = 1:trials
                                                    for i = 1:trials
 A = sparse(n, n);
                                                       A = sparse(n, n);
                                                       for k = 1:numnonzeroentries
 for k = 1:numnonzeroentries
   i = randi(n); j = randi(n);
                                                        i = randi(n); j = randi(n);
   A(i, j) = randn(1,1);
                                                        A(i, j) = randn(1,1);
 b = randn(n, 1);
                                                       b = randn(n, 1);
  starttime = cputime;
                                                       starttime = cputime;
  x = A \setminus b;
                                                       x = A \setminus b;
  endtime = cputime;
                                                       endtime = cputime;
  totaltime = totaltime + (endtime - starttime);
                                                       totaltime = totaltime + (endtime - starttime);
t1000 = totaltime / trials
                                                     t2000 = totaltime / trials
                                                          alpha =
          alpha = log2(t2000/t1000)
          c = t1000 / 1000^alpha
                                                              2.6274
          n24 = (86400 / c)^(1/alpha)
                                                          c =
                                                             8.3090e-10
                                                             2.1627e+05 ~
```

You may have noticed that the running time is highly variable. Running for more trials would help a little bit, but probably a different problem would give a more useful timing result.

```
import time
import numpy as np
                                             import numpy as np
n = 2000
                                             n = 1000
                                             trials = 10
trials = 10
totaltime = 0
                                             totaltime = 0
for in range(trials):
                                             for _ in range(trials):
 A = np.random.randn(n, n)
                                              A = np.random.randn(n, n)
 B = np.random.randn(n, n)
                                               x = np.random.randn(n)
  start = time.perf_counter()
                                               b = A.dot(x)
 C = A.dot(B) # or np.dot(A,B)
                                               start = time.perf_counter()
                # NOT A * B!
                                               np.linalg.solve(A, b)
  totaltime += time.perf_counter() - start totaltime += time.perf_counter() - start
                                             averagetime = totaltime / trials
averagetime = totaltime / trials
                                             averagetime
averagetime
    import time
    import numpy as np
    import scipy.sparse
    from scipy.sparse.linalg import spsolve, bicg
    n = 500
```

```
import time
import numpy as np
import scipy.sparse
from scipy.sparse.linalg import spsolve, bicg
total = 0
trials = 10
for _ in range(trials):
  sparsity = 10
  rows = np.random.randint(n, size=sparsity*n)
  cols = np.random.randint(n, size=sparsity*n)
  data = np.random.randn(sparsity*n)
 A = scipy.sparse.csr_matrix((data, (rows,cols)), shape=(n,n))
#A += 1000* scipy.sparse.eye(n) ## uncomment this to improve the condition no
  #A = A.toarray()
                                     ## uncomment this to use dense matrix
  x = np.random.randn(n,1)
  b = A.dot(x)
  start = time.perf_counter()
  c = bicg(A,b)
  end = time.perf_counter()
  total += end - start
total
```