

1. Assuming that each no. is stored as $0.abcd \times 10^x$,
 & two digits are stored after decimal. i.e. $0.ab \times 10^x$.

$$\left[\begin{array}{ccc|c} 0.4 \times 10^1 & 0.1 \times 10^1 & 0.2 \times 10^1 & 0.9 \times 10^1 \\ 0.2 \times 10^1 & 0.4 \times 10^1 & -0.1 \times 10^1 & -0.5 \times 10^1 \\ 0.1 \times 10^1 & 0.1 \times 10^1 & -0.3 \times 10^1 & -0.9 \times 10^1 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 1 & 0.25 & 0.5 & 2.25 \\ - & - & - & - \\ - & - & - & - \end{array} \right]$$

$$\downarrow \sim$$

$$\left[\begin{array}{ccc|c} 1 & 0.25 & 0.5 & 2.2 \\ 0.2 & 4 & -1 & -5 \\ 1 & 1 & -3 & -9 \end{array} \right] \quad \left(\begin{array}{l} 2.25 \sim 0.225 \times 10^1 \\ \sim 0.22 \times 10^1 \\ \sim 2.2 \end{array} \right)$$

$$\downarrow$$

$$\left[\begin{array}{ccc|c} 1 & 0.25 & 0.5 & 2.2 \\ 0 & 3.5 & -2 & -9.4 \\ 0 & 0.75 & -3.5 & -11.2 \end{array} \right]$$

$$\downarrow \sim$$

$$\left[\begin{array}{ccc|c} 1 & 0.25 & 0.5 & 2.2 \\ 0 & 3.5 & -2 & -9.4 \\ 0 & 0.75 & -3.5 & -11 \end{array} \right] \quad \left(\begin{array}{l} 11.2 \sim 0.112 \times 10^2 \\ \sim 0.11 \times 10^2 \\ \sim 11 \end{array} \right)$$

$$\downarrow$$

$$\left[\begin{array}{ccc|c} 1 & 0.25 & 0.5 & 2.2 \\ 0 & 1 & -0.57 & -2.69 \\ 0 & 0.75 & -3.5 & -11 \end{array} \right]$$

$$\downarrow \sim$$

$$\left[\begin{array}{ccc|c} 1 & 0.25 & 0.5 & 2.2 \\ 0 & 1 & -0.57 & -2.7 \\ 0 & 0.75 & -3.5 & -11 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 1 & 0.25 & 0.5 & 2.2 \\ 0 & 1 & -0.57 & -2.7 \\ 0 & 0 & -3.1 & -9 \end{array} \right]$$

$$\therefore x_3 = \frac{9}{3.1} = 2.9 = 0.29 \times 10^1$$

$$x_2 = -2.7 + 0.57 \times 2.9 = -1.04 \approx -0.10 \times 10^1$$

$$x_1 = 2.2 - 0.5 \times 2.9 + 0.25 \times 1 = \cancel{0.5 \times 10^0} \\ = 1.00 = 0.10 \times 10^1$$

2.

- (a) `numpy.fft.fft(Python)`
- (b) `numpy.linalg.qr(Python)`
- (c) `numpy.random.lognormal (Python)`
- (d) ~~`numpy`~~ (`gsl_odeiv2.h`) with `gsl_odeiv2_step_rk8pd`
or ~~`np`~~ or `scipy.integrate.solve_ivp(method='DOP853')` (C)
- (e) `numpy.linalg.svd (Python)`
- (f) ~~`gsl_odeiv2.h`~~ `scipy.stats (Python)`, `gsl_randist.h (C)`
- (g) (`gsl_odeiv2.h`) • `gsl_odeiv2_control` function.
- (h) `scikit monaco (Python)`

<code>gsl_monte_plain.h</code>	}	C
<code>gsl_monte_mixed.h</code>		
<code>gsl_monte_vegas.h</code>		
- (i) `scipy.integrate.solve_bvp (Python)`
- (j) `numpy.linalg.eig (Python)`

$$3. A = \begin{bmatrix} a_{11} & a_{12} & 0 & 0 & 0 & \dots & \dots \\ a_{21} & a_{22} & a_{23} & 0 & 0 & \dots & \dots \\ 0 & a_{32} & a_{33} & a_{34} & 0 & \dots & \dots \\ 0 & 0 & a_{43} & a_{44} & a_{45} & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & a_{n-1,n-2} & a_{n-1,n-1} & a_{n-1,n} \\ 0 & 0 & \dots & 0 & a_{nn-1} & a_{nn} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}$$

Steps:

$R_1/a_{11} \rightarrow 3$ ~~step~~ operations.

$-(R_1 \times a_{21}) + R_2 \rightarrow 3$ 'x' & 3 '-'.
 $\therefore 9$ ~~operations~~ operations followed by $\frac{R_2}{a_{22}} \rightarrow 3 \rightarrow 12$ operations for row 1 & 2

After these problem becomes:

$$\begin{bmatrix} 1 & a_{12}' & 0 & 0 & \dots & b_1' \\ 0 & 1 & a_{23}' & \dots & \dots & b_2' \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{bmatrix}$$

Similarly for row 2 & 3 - 12 operations.

Thus $1^2 \times (n-1)$ such operations.

$$\begin{bmatrix} 1 & a_{12}' & 0 & 0 & \dots & b_1' \\ 0 & 1 & a_{23}' & 0 & \dots & b_2' \\ 0 & 0 & 1 & a_{34}' & \dots & b_3' \\ 0 & 0 & 0 & 1 & a_{45}' & b_4' \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & a_{n-1,n}' & b_{n-1}' \\ 0 & 0 & \dots & 0 & a_{nn}' & b_n' \end{bmatrix}$$

After which matrix becomes:
Backward subⁿ:

Then $x_n = b_n' / a_n'$

for each x_i ,

$$x_i = b_i' - a_{i-1,i}' \times x_{i-1}$$

\downarrow
 2 operations.

Thus total no. of operation: $12(n-1) + 2(n-1) + 1$
 $= 14(n-1) + 1$ (roughly, there might be some error in counting but the order of magnitude of no. of operations is intact)
 $= O(n)$

$12(n-1) \rightarrow$ row operations

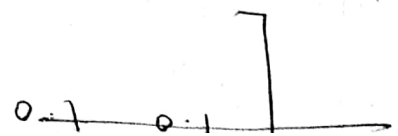
$2(n-1) \rightarrow$ backward substitution.

$$4(c): k_{\max} = 2\pi \left(1 - \frac{2}{1024}\right) \approx 3.1354$$

$$k_{\min} = -\pi \approx -3.1416$$

$$x = 0, 1, 2, \dots, 1024$$

$$dx = 1$$



4(e):

It was shown in class that Power spectrum of a data is the fourier transform of its correlation function. (time averaged).

$$\text{Correlation function: } E[f_x(x_1) f_x(x_2)] = \begin{cases} R(x_1, x_2) & \text{if } |x_1|, |x_2| < X \\ 0, & \text{otherwise.} \end{cases}$$

for unifor random distribution,

$f(x_1)$ & $f(x_2)$ are independent.

$$\begin{aligned} \text{Thus } E[f_x(x_1) f_x(x_2)] &= E(f(x_1)) E(f(x_2)) \\ &= 0.5 \times 0.5 \quad (f \text{ is uniform bet}^n (0,1)) \\ &= 0.25 \\ &= \text{const} \\ &\text{for any } x_1, x_2. \end{aligned}$$

Thus time-average is also independent of $\tau = (x_1 - x_2)$

Now, fourier transform of constant function is a delta-function, which is what I get in Python.

Another way to see is that power spectrum qualitatively gives the contribution of e^{ikx} to $f(x)$. But here since $f(x)$ is uniformly distributed only at $k=0$ ~~contribute~~ & hence power spectrum is a delta function.

5. Criteria to choose library:

- i) Price^{& source code access}: It might sound weird, but open-source libraries are always preferred above licensed ~~one~~ ones. They serve 2 advantages: They are free & they can be ~~modie~~ modified as per our requirement.
- ii) Versatility: I would like to choose a library which work for quite general kind of problems. e.g. If I want a library for fourier transform, I expect it to work for multidimensions. ~~I can~~
- iii) Storage: The functions in library ~~is~~ generate various intermediate arrays & variables which need storage. A good library ~~have~~ require less storage for such arrays or variables.
- iv) Speed: This is an obvious ~~cho~~ criteria. The faster the functions in library, the better.
- v) Compatibility with language:- The library must be compatible with the language in which I do most of my coding work.

$$\begin{aligned}
 6. \quad \frac{d}{dx} (y_1 + 2y_2) &= 32y_1 - 132y_1 + 66y_2 - 266y_2 \\
 &= -100y_1 - 200y_2 \\
 &= -100(y_1 + 2y_2)
 \end{aligned}$$

$$\begin{aligned}
 \therefore y_1 + 2y_2 &= e^{-100x} (y_1 + 2y_2)(0) \\
 &= e^{-100x}
 \end{aligned}$$

Thus $y_1 + 2y_2 \approx 0$ Since e^{-100x} rapidly goes to 0.

Thus $y_1 \approx -2y_2$, which is also observed in the obtained solution.

$$7. \quad x_{i+1} = (x_i \times a + c) \% m$$

$a \rightarrow$ multiplier

$c \rightarrow$ increment

$m \rightarrow$ modulus.

Clearly all nos. lie betⁿ 0 to $m-1$.

$x_0 \rightarrow$ seed.

i) seed repeats:

e.g. $a = 17$

$c = 0$

$m = 32$

$x_0 = 1$

Sequence: [1 17 33 49 65 81 97 113 (1) 17 ...]

Thus we see that the seed reappears.

ii) Seed does not ~~re~~ reappear:

This can occur if we get $x=0$ somewhere along the chain, in that case the sequence gets stuck to c .

Or if some value other than seed appears again & hence ~~seed~~ ^{seed} cannot appear again.

1st kind of e.g. is $a=2, c=0, m=2^y, x_0=1$.

In that case $x_0=1, x_1 = (2 \times 1) \% 2^y = 2, x_2 = 4, x_3 = 8$

... $x_y = 0, x_{y+1} = 0 \dots$

Thus we never get 1 again. Though this is very bad generator.

Another e.g. is $a = \text{even}$
 $c = \text{even}$
 $m = \text{even}$

$k, x_0 = \text{odd}$

In that case $x_1 = (\text{even} \times \text{odd} + \text{even}) \bmod \text{even}$
 $= \text{even}$

Thus every new no. will be even. Thus
seed never appear again.