

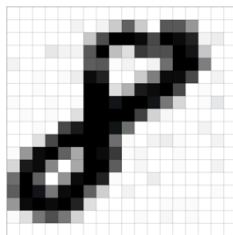
# Foundational Concepts in Numerical Linear Algebra

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資料科學計算, Spring 2018



# Matrix Representations

## Image Data as a Matrix

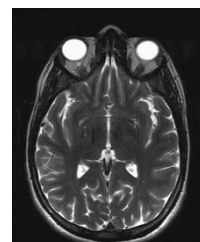
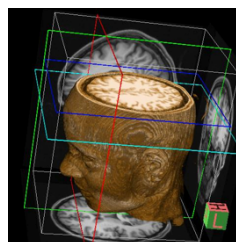


```

0 0 0 0 1 12 0 1 1 3 3 137 37 0 152 147 34 0 0 0 0
0 0 0 1 0 1 1 10 258 255 235 262 255 238 206 11 13 0
0 0 1 5 9 9 159 21 4 21 384 152 254 253 23 0 0 0
0 0 0 0 0 0 8 165 146 1 8 0 11 124 253 165 10 0 0
0 3 0 4 15 258 256 0 8 38 169 147 349 389 9 11 0
0 2 0 0 8 253 253 0 0 254 254 255 384 5 4 0 0
0 2 0 0 2 255 255 255 255 255 255 255 255 255 255
0 4 0 169 255 255 255 255 255 255 255 255 255 255
0 2 162 255 255 255 255 255 255 255 255 255 255
3 79 242 251 34 66 255 255 7 8 0 0 5 0 0 0 0 0
23 221 58 6 0 255 255 144 0 0 0 0 0 7 0 0 11 0
125 253 5 9 87 144 255 255 0 0 13 0 1 0 1 0 0
23 234 255 255 255 341 34 11 0 1 0 0 0 1 3 0
23 234 255 255 255 255 255 255 255 255 255 255
5 23 112 157 114 32 0 0 0 0 0 0 0 7 0 0 0 0

```

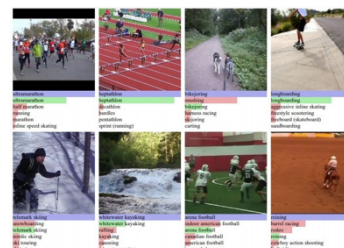
## MRI Images as a 3D Matrix or Multiple 2D Matrices

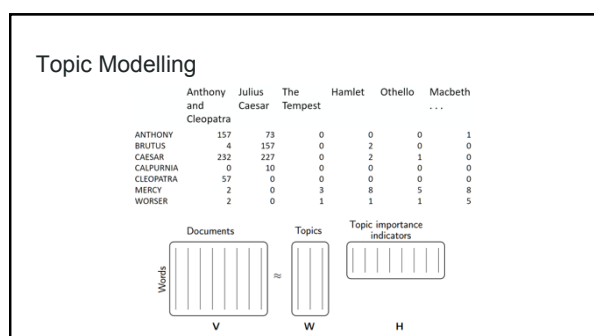
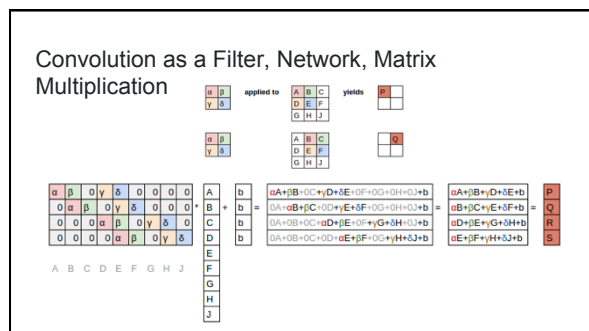
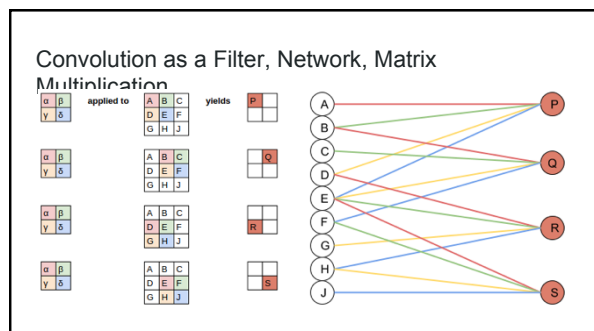


# The Key Concepts of Linear Algebra

- From "Numbers" to "Vectors (can be matrices or functions too)" to "Spaces"
- The four subspaces

## Convolution in CNN





# Floating Point Numbers

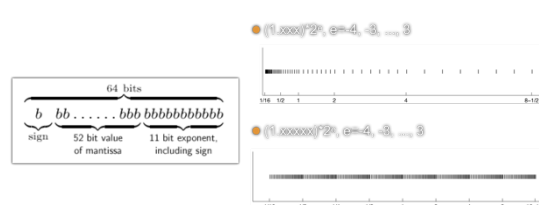
## Floating Point Numbers

$k$	$2^{-k}$	$b_k$	$r_k = r_{k-1} - b_k 2^{-k}$
0	NA	NA	0.1
1	0.5	0	0.1
2	0.25	0	0.1
3	0.125	0	0.1
4	0.0625	1	0.1 - 0.0625 = 0.0375
5	0.03125	1	0.0375 - 0.03125 = 0.00625
6	0.015625	0	0.00625
7	0.0078125	0	0.00625
8	0.00390625	1	0.00625 - 0.00390625 = 0.00234375
9	0.001953125	1	0.00234375 - 0.001953125 = 0.000390625
10	0.0009765625	0	0.000390625
:	:	:	:

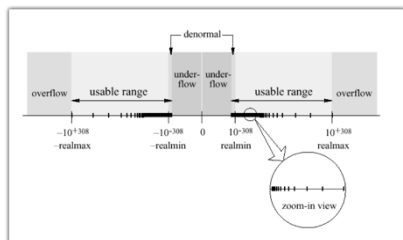
Therefore, the binary mantissa for 0.1 is  $(00011\,0011\,\dots)_2$ .

Therefore, the binary mantissa for 0.1 is  $(00011\ 0011\ \dots)_2$ .

## Floating Point Numbers



## Floating Point Numbers



## Floating Point Numbers

$$f(x) = \frac{1 - \cos x}{x^2}, \quad x \neq 0$$

10 decimal digits, and it used rounded arithmetic.

$x$	Computed $f(x)$	True $f(x)$
0.1	0.4995834700	0.4995834722
0.01	0.4999960000	0.4999958333
0.001	0.5000000000	0.499999583
0.0001	0.5000000000	0.499999996
0.00001	0.0	0.5000000000

## Floating Point Numbers

- 兩相近數相減，導致有效位數減損誤差 (loss of significance error)
- 數學上相等，計算效益不同

$$\begin{aligned} f(x) &= \frac{1 - \cos x}{x^2} = \frac{2 \sin^2(x/2)}{x^2} \\ &= \frac{1}{2} \left[ \frac{\sin(x/2)}{x/2} \right]^2 \end{aligned}$$

## Ariane 5 Explosion



## Ariane 5 Explosion

- 慣性參考系統軟體出錯
- 紀錄側向速度的64位元浮點數，在慣性參考系統軟體中，被轉換成16位元的整數
- 但卻超過其所能表示的最大數 (32,767)
- 慣性參考系統因此被認定故障，向主電腦送出錯誤信號，並自動關機
- 控制火箭主電腦，錯把錯誤訊息當成火箭 當時的火箭狀況參數，做出不必要的方向修正與旋轉
- 推進器與火箭因而被空氣動力裂解終於導致安全系統啟動，自動引爆

## Quiz

Estimate the size of a 3D MRI head image.

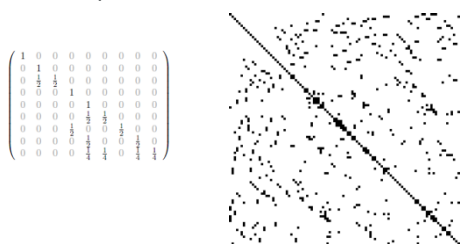
## Memory Usage

### Dense and Sparse Matrix

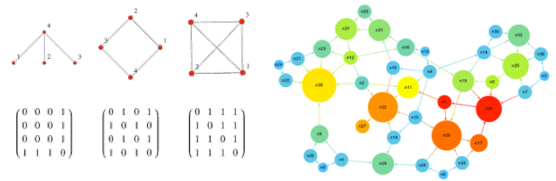
$\alpha$	$\beta$	0	$\gamma$	$\delta$	0	0	0	0
0	$\alpha$	$\beta$	0	$\gamma$	$\delta$	0	0	0
0	0	0	$\alpha$	$\beta$	0	$\gamma$	$\delta$	0
0	0	0	0	$\alpha$	$\beta$	0	$\gamma$	$\delta$

 $\begin{matrix} A \\ B \\ C \\ D \\ E \\ F \\ G \\ H \\ J \end{matrix}$  $\begin{matrix} b \\ b \\ b \\ b \end{matrix}$  $\begin{matrix} A \\ B \\ C \\ D \\ E \\ F \\ G \\ H \\ J \end{matrix}$  $\begin{matrix} b \\ b \\ b \\ b \end{matrix}$  $\begin{matrix} \alpha A + \beta B + 0C + \gamma D + \delta E + 0F + 0G + 0H + 0J + b \\ \alpha A + \beta B + 0C + 0D + \gamma E + \delta F + 0G + 0H + 0J + b \\ \alpha A + \beta B + 0C + 0D + 0E + \delta F + \gamma G + \delta H + 0J + b \\ \alpha A + \beta B + 0C + 0D + \alpha E + \beta F + 0G + \gamma H + \delta J + b \end{matrix}$  $\begin{matrix} \alpha A + \beta B + \gamma D + \delta E + b \\ \alpha B + \beta C + \gamma E + \delta F + b \\ \alpha D + \beta E + \gamma G + \delta H + b \\ \alpha E + \beta F + \gamma H + \delta J + b \end{matrix}$  $\begin{matrix} P \\ Q \\ R \\ S \end{matrix}$

### Dense and Sparse Matrix

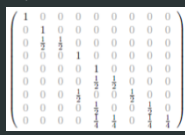


### Graphic Matrix



## Quiz

How do you store the following sparse matrix with as less of memory as possible?

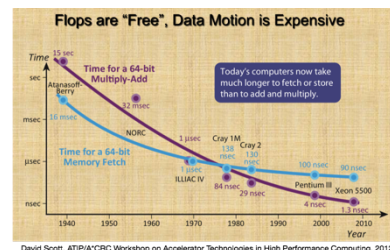


## Performance

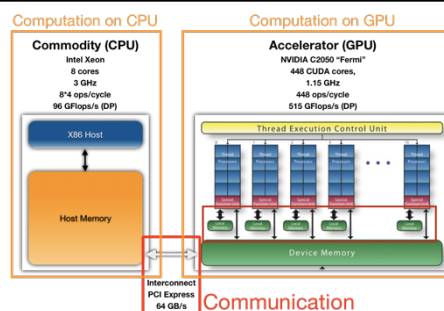
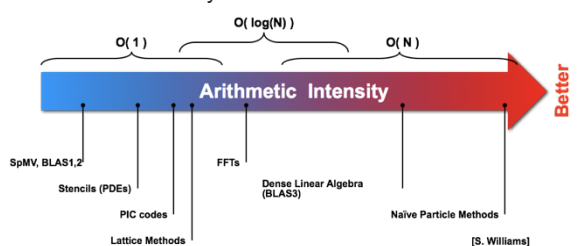
## Complexity

- Computational complexity
  - Big O notation
  - Usually in terms of number of rows (m) and number of columns (n)
  - Basic Linear Algebra Subroutines: BLAS 1, BLAS 2, BLAS 3
  - What about sparse matrix?
- Communication complexity
  - Getting more and more important

## Computation and Communication

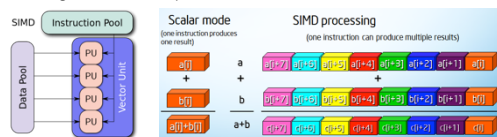


## Arithmetic Intensity



## Vectorization

- SIMD: Single instruction, multiple data

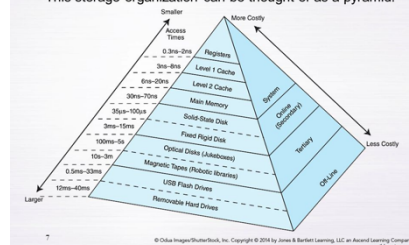


- Will use vectorized operations in libraries (e.g. numpy, which in turn rely on specially tuned vectorized low level linear algebra APIs in particular, BLAS, and LAPACK)

## Locality

### 6.3 The Memory Hierarchy

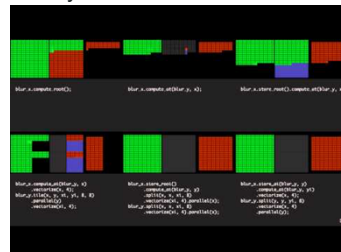
- This storage organization can be thought of as a pyramid:



## Locality

- To do computation required at that time, rather than having to load it multiple times each time we need it. Try to get higher arithmetics intensity (i.e. comp/ access) 每拿一次資料，盡量多算一些。
- To access data items that are stored next to each other, so try to always use any data stored nearby that we know we'll need soon.  
近期需要的資料，儘量放在附近記憶體
- Locality Numbers Every Programmer Should Know
  - [https://people.eecs.berkeley.edu/~rcs/research/interactive\\_latency.html](https://people.eecs.berkeley.edu/~rcs/research/interactive_latency.html)

## Locality



- Locality is hard.
  - Potential trade-offs:
    - redundant computation to save memory bandwidth
    - sacrificing parallelism to get better reuse

## Temporaries

- Temporary variables in RAM can significantly slow down the computation, comparing with in cache
- Numpy generally creates temporaries for every single operation or function it does.
  - E.g.  $a=b-c^2+\ln(d)$  will create four temporaries (since there are four operations and functions)

## Team Time

Complexity in Big O Notations  
(Using not-boring math to measure code's efficiency)

<https://goo.gl/yvKkKE>

