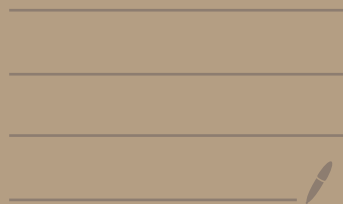


Week 12

Applications of Linear Algebra



Spectral Theorem

- Recall :
- $A \in \mathbb{R}^{n \times n}$ is symmetric if $A^T = A$
 - $Q \in \mathbb{R}^{n \times n}$ is orthogonal if $Q^T Q = I$

Th (Spectral Theorem for symmetric matrices)

If $A \in \mathbb{R}^{n \times n}$ is symmetric, then :

- All eigenvalues of A are real
- Eigenvectors of A corresponding to distinct eigenvalues are orthogonal.
- A can be diagonalized by an orthogonal matrix, i.e., there exists an orthogonal matrix $Q \in \mathbb{R}^{n \times n}$ s.t.

$$Q^T A Q = D,$$

where D is diagonal.

Remark To diagonalize a symmetric A by an orthogonal

- matrix Q :
- find a basis for each eigenspace and use Gram-Schmidt to make it orthonormal
 - by Spectral Th, orthonormal bases of all eigenspaces jointly form an orthonormal basis for \mathbb{R}^n
 - form the matrix Q using the orthonormal basis vectors as columns.

Applications of Linear Algebra

① Solving linear recursions / difference equations

Example: Fibonacci sequence

$$F_0 = 0, F_1 = 1, F_{n+2} = F_n + F_{n+1}$$

$$\rightsquigarrow 0, 1, 1, 2, 3, 5, 8, \dots \quad F_n = ?$$

$$\begin{pmatrix} F_{n+1} \\ F_n \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} F_n \\ F_{n-1} \end{pmatrix} = \dots = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix}^n \begin{pmatrix} 1 \\ 0 \end{pmatrix}$$

$$\text{SHIFT} \rightarrow \sigma \underline{x} = A \underline{x}, \quad A = \begin{pmatrix} 1 & 1 \\ 1 & 0 \end{pmatrix};$$

$$p_A(\lambda) = \lambda^2 - \lambda - 1, \quad \text{eigenvalues } \lambda_1 = \frac{1+\sqrt{5}}{2}, \lambda_2 = \frac{1-\sqrt{5}}{2}$$

$$\text{eigenvectors } \underline{v}_1 = \begin{pmatrix} \lambda_1 \\ 1 \end{pmatrix}, \underline{v}_2 = \begin{pmatrix} \lambda_2 \\ 1 \end{pmatrix}$$

Diagonalize A :

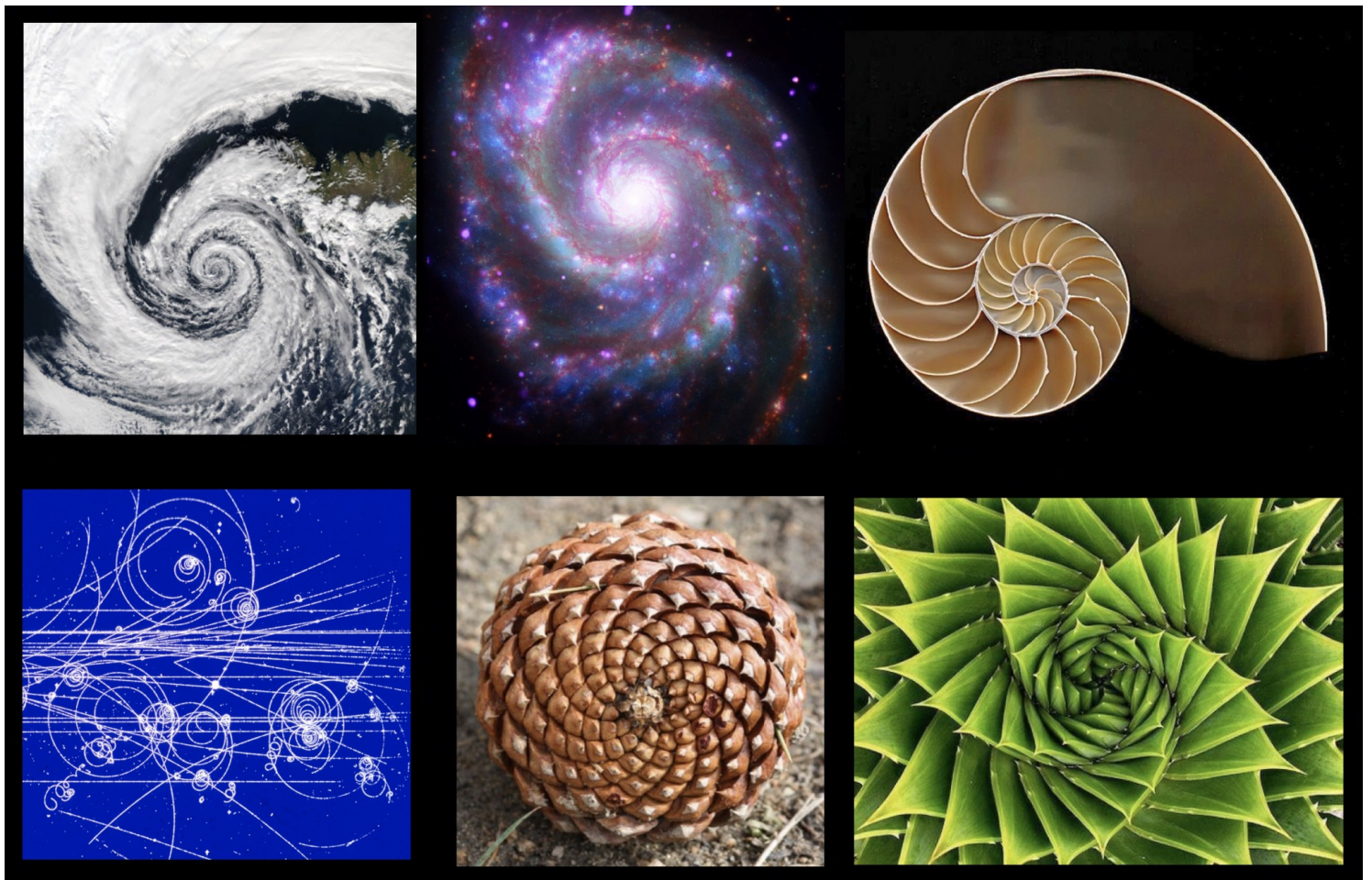
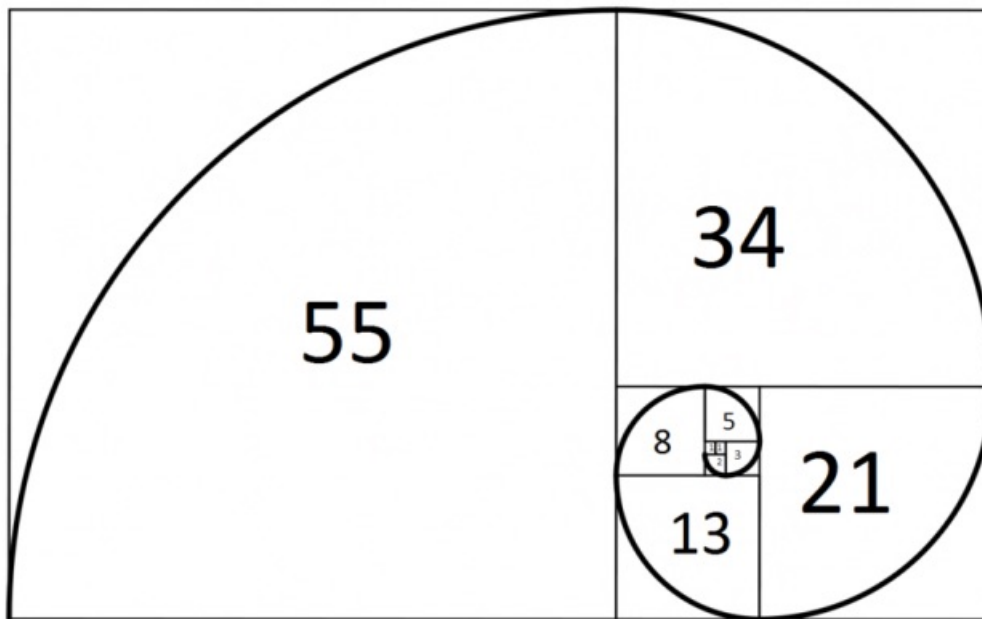
$$P = \begin{pmatrix} \lambda_1 & \lambda_2 \\ 1 & 1 \end{pmatrix}, \quad P^{-1}AP = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix}$$

$$\Rightarrow A = P \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} P^{-1}, \quad P^{-1} = \frac{1}{\lambda_1 - \lambda_2} \begin{pmatrix} 1 & -\lambda_2 \\ -1 & \lambda_1 \end{pmatrix}$$

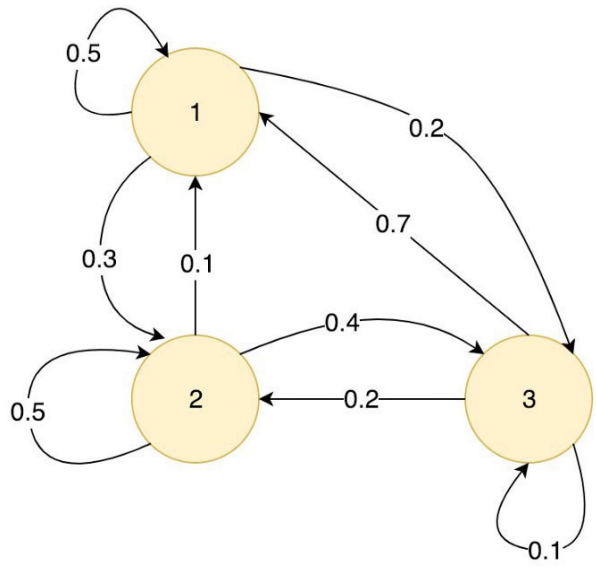
$$\Rightarrow A^n = P \begin{pmatrix} \lambda_1^n & 0 \\ 0 & \lambda_2^n \end{pmatrix} P^{-1} = \frac{1}{\lambda_1 - \lambda_2} \begin{pmatrix} \lambda_1 & \lambda_2 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \lambda_1^n & 0 \\ 0 & \lambda_2^n \end{pmatrix} \begin{pmatrix} 1 & -\lambda_2 \\ -1 & \lambda_1 \end{pmatrix}$$

$$\text{and } \begin{pmatrix} F_{n+1} \\ F_n \end{pmatrix} = A^n \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \boxed{F_n = \frac{\lambda_1^n - \lambda_2^n}{\lambda_1 - \lambda_2}}$$

Golden ratio in nature :



(2) Markov Chains \rightarrow finance



Markov process

	S_1	S_2	S_3
S_1	0.5	0.1	0.7
S_2	0.3	0.5	0.2
S_3	0.2	0.4	0.1

Transition matrix

③ Linear differential equations

$$\underline{y}' = A \underline{y}, \text{ where } \underline{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}, \underline{y}' = \begin{pmatrix} y_1' \\ \vdots \\ y_n' \end{pmatrix}$$

The solution set consists of

$y \in (C^1(\mathbb{R}))^n$ that satisfy the equation, and it has a structure of a vector space:

If $\underline{u}' = A \underline{u}$ and $\underline{v}' = A \underline{v}$, then

$$(\underline{u} + \underline{v})' = \underline{u}' + \underline{v}' = A \underline{u} + A \underline{v} = A (\underline{u} + \underline{v})$$

$$\text{and } (\lambda \underline{u})' = \lambda \underline{u}' = \lambda A \underline{u} = A (\lambda \underline{u}).$$

Ex $y_1' = 2y_1 - 3y_2$ \rightsquigarrow $\underline{y}' = \begin{pmatrix} 2 & -3 \\ 1 & -2 \end{pmatrix} \underline{y}$
 $y_2' = y_1 - 2y_2$

$A = \begin{pmatrix} 2 & -3 \\ 1 & -2 \end{pmatrix}$ is diagonalizable: $A = P D P^{-1}$, $P = \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix}$, $D = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}$

Change of variables: $\underline{y} = P \underline{w}$ (i.e. $\underline{w} = P^{-1} \underline{y}$)

$$\underline{y}' = P D P^{-1} \underline{y} \Rightarrow P^{-1} \underline{y}' = D P^{-1} \underline{y} \Rightarrow (\underline{P}^{-1} \underline{y})' = D (\underline{P}^{-1} \underline{y})$$

$$\Rightarrow \underline{w}' = D \underline{w}, \text{ i.e. } \begin{aligned} w_1' &= w_1 \\ w_2' &= -w_2 \end{aligned}$$

$$\Rightarrow \begin{aligned} w_1(t) &= C_1 e^t \\ w_2(t) &= C_2 e^{-t} \end{aligned}$$

$$\Rightarrow \underline{y} = P \underline{w} = \begin{pmatrix} 3 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} C_1 e^t \\ C_2 e^{-t} \end{pmatrix}$$

$$\Rightarrow y_1 = 3C_1 e^t + C_2 e^{-t}$$

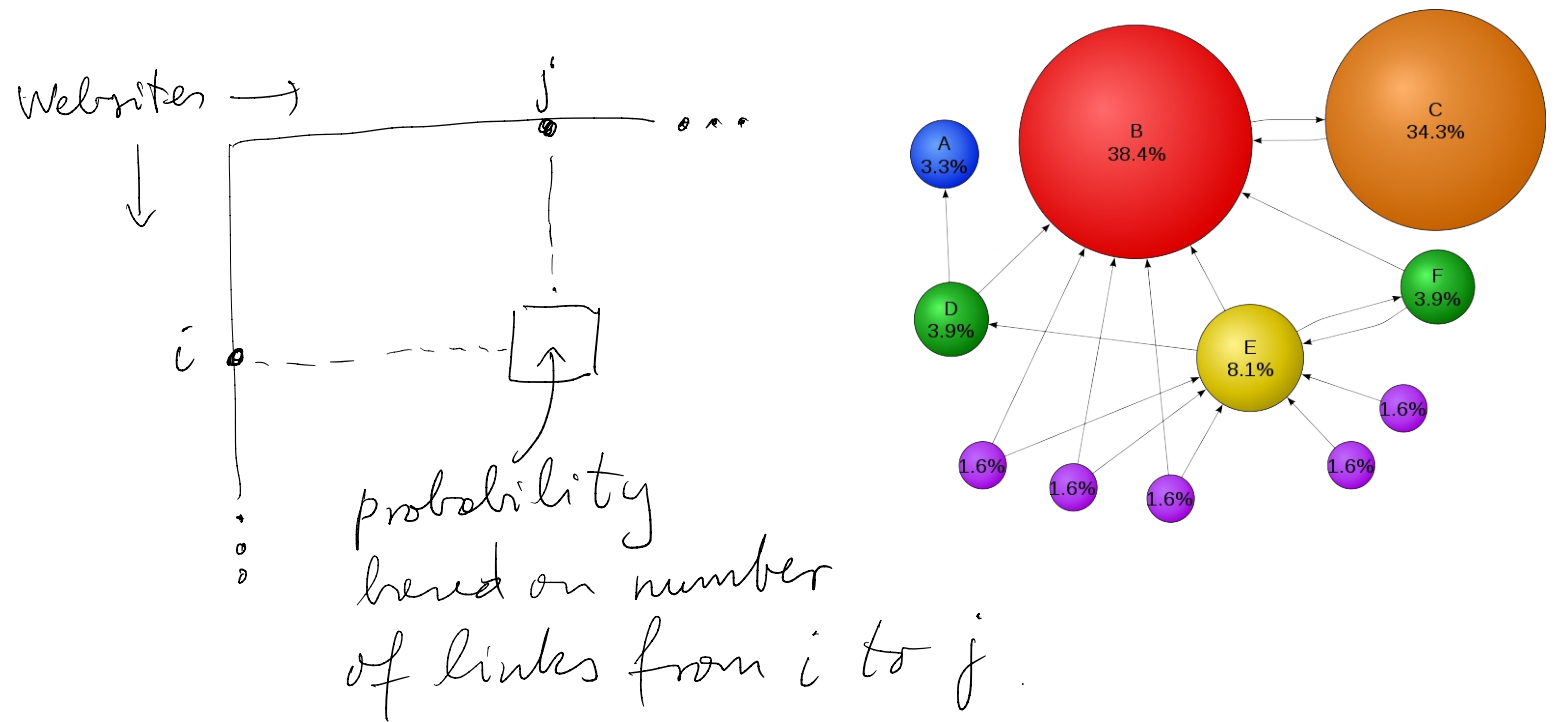
$$y_2 = C_1 e^t + C_2 e^{-t}.$$

"general" solution

Solution set has dimension 2.

④ Google PageRank algorithm

Incidence matrix of the internet



Calculate approximate eigenvectors.

- ⑤ Networks :
- electrical
 - social
 - transport
 - disease transmission

⑥ Image / signal processing

- ⑦ 3D Computer Graphics :
- CGI effects in films
 - computer games
 - design

Work in \mathbb{R}^3

- Geometric transformations :
- rotations
 - reflections
 - scaling, etc

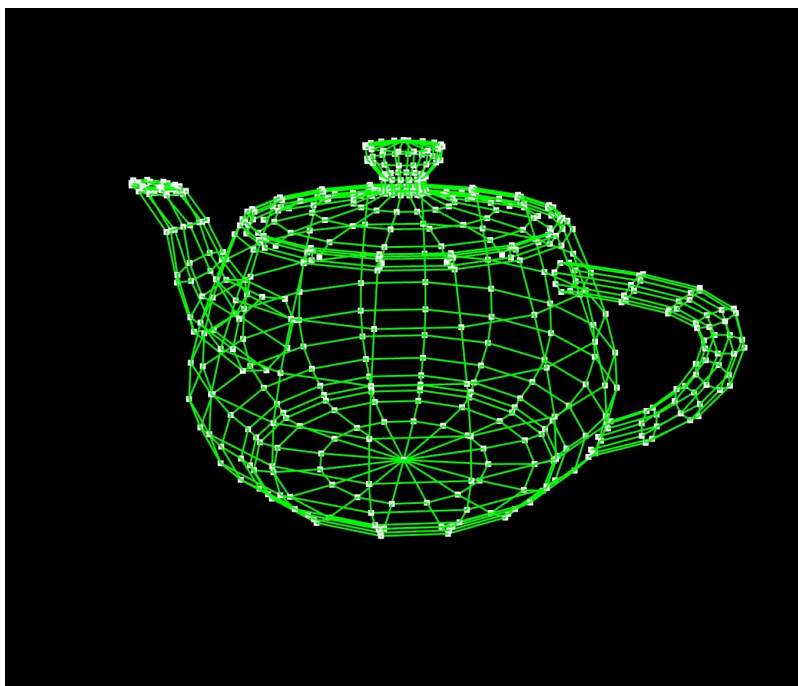
are in fact linear transformations $\mathbb{R}^3 \rightarrow \mathbb{R}^3$

$$\underline{x} \mapsto A \underline{x}$$

for a suitable A :

- rotation if A is orthogonal with $\det(A) = 1$
- reflection — — — — — $\det(A) = -1$

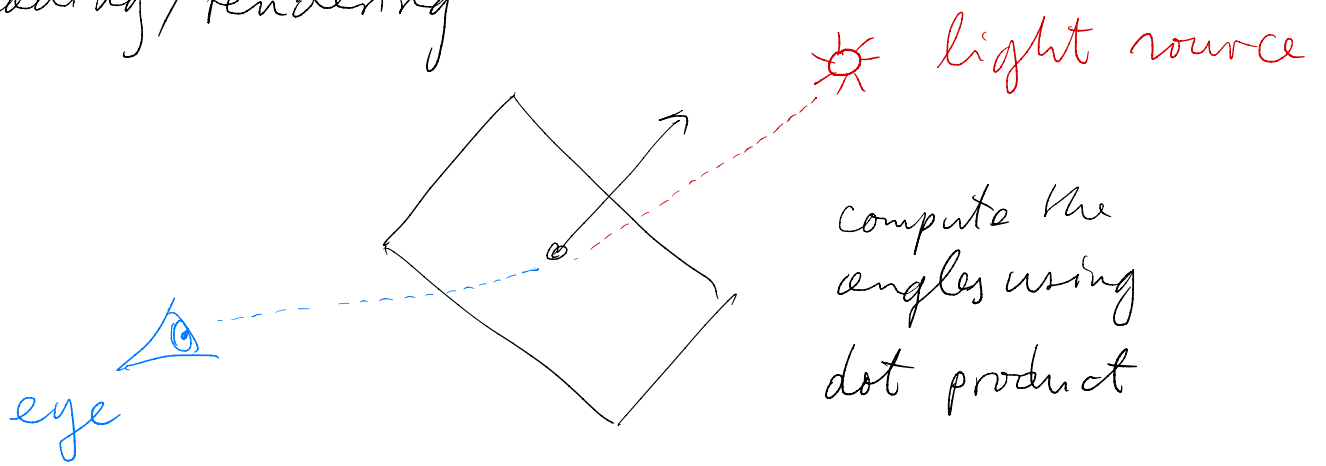
Representing shapes:



Points are represented
by $\underline{x}_1, \underline{x}_2, \dots, \underline{x}_n \in \mathbb{R}^3$

↪ matrix of coordinates
 $X \in \mathbb{R}^{3 \times n}$

Shading/rendering



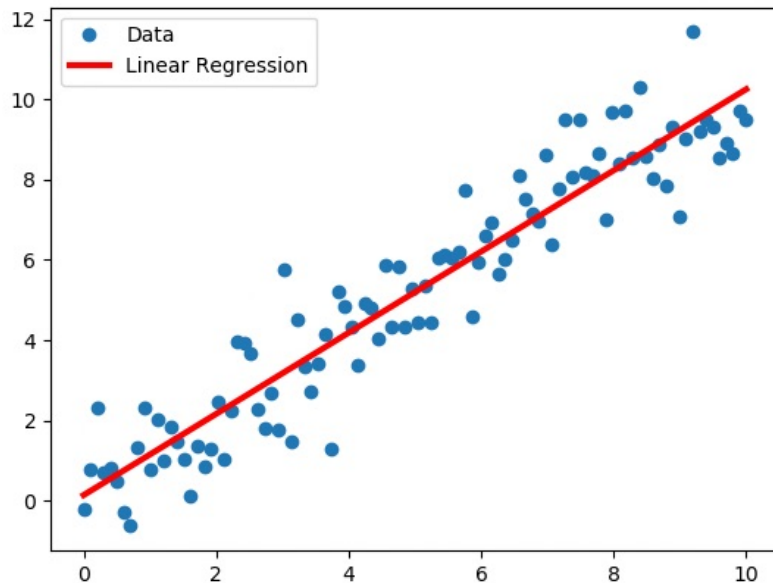
Animate ; Perform a transformation

$$X \mapsto A X$$

and draw/render the new shape incrementally (A 's vary in time).

⑧ Machine learning - neural networks

⑨ Data science / analysis



Linear regression \leftrightarrow least squares method

⑩ Coding Theory (QMUL module)

- linear codes are vector spaces

over finite fields $\mathbb{F}_p = \mathbb{Z}/p\mathbb{Z}$

⑪ Cryptography (QMUL module)