

Quantum versus Classical Measures of Complexity on Classical Information

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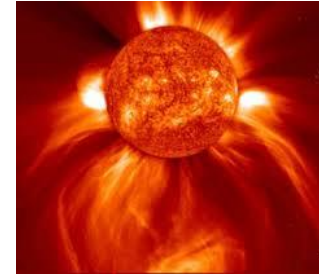
Complexity Institute
Nanyang Technological University



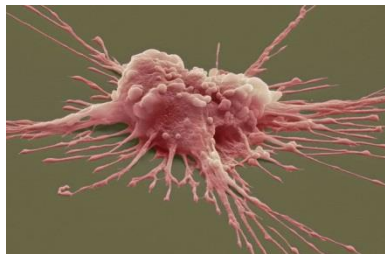
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Astronomical Scale



Macroscopic Scale



Mesoscopic Scale



What is Complexity?

- Complexity is related to “Pattern” and “Organization”.

*Nature inherently organizes;
Pattern is the fabric of Life.*

James P. Crutchfield

- There are two extreme forms of Pattern: generated by a Clock and a Coin Flip.
- The former encapsulates the notion of Determinism, while the latter Randomness.
- Complexity is said to lie between these extremes.

Can Complexity be Measured?

- To measure Complexity means to measure a system's structural organization. How can that be done?
- Conventional Measures:
 - Difficulty in Description (in bits) – Entropy; Kolmogorov-Chaitin Complexity; Fractal Dimension.
 - Difficulty in Creation (in time, energy etc.) – Computational Complexity; Logical Depth; Thermodynamic Depth.
 - Degree of organization – Mutual Information; Topological ϵ -machine; Sophistication.

Topological \mathcal{E} -Machine

1. *Optimal* Predictor of the System's Process
2. *Minimal* Representation – Ockham's Razor
3. It is *Unique*.
4. It gives rise to a new measure of Complexity known as *Statistical Complexity* to account for the degree of organization.
5. The Statistical Complexity has an essential kind of *Representational Independence*.

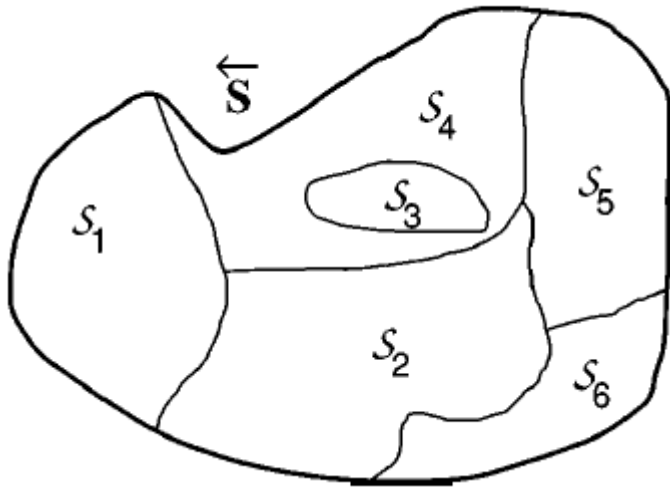
Concept of \mathcal{E} -Machine

Symbolic Sequences: $\dots S_{-2}S_{-1}S_0S_1 \dots$

Futures : $\vec{S}_t = S_t S_{t+1} S_{t+2} \dots$

Past : $\overleftarrow{S}_t = \dots S_{t-3} S_{t-2} S_{t-1}$

Partitioning the set $\overleftarrow{\mathcal{S}}$ of all histories into causal states S_i .



- Within each partition, we have

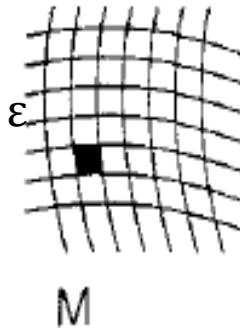
$$P(\vec{S} | \overleftarrow{s}) = P(\vec{S} | \overleftarrow{s}') \quad (1)$$

where \overleftarrow{s} and \overleftarrow{s}' are two different individual histories in the same partition.

- Equation (1) is related to subtree similarity discussed later.

\mathcal{E} Machine - Preliminary

- In physical processes, measurements are taken as time evolution of the system.
- The time series is converted into a sequence of symbols $s = s_1, s_2, \dots, s_i, \dots$ at time interval τ .

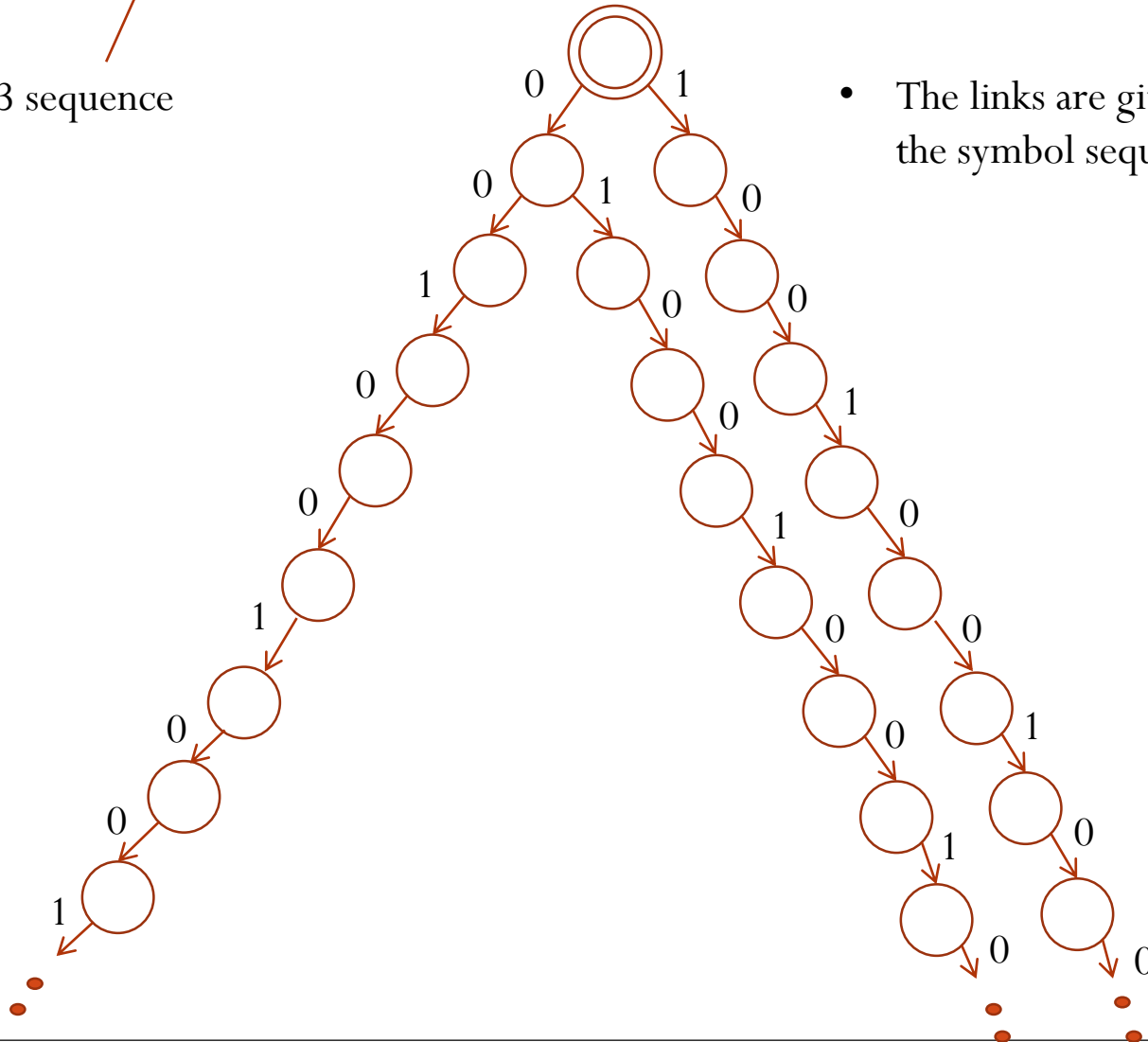


- Measurement phase space M is segmented or partitioned into cells of size ϵ .
- Each cell can be assigned a label leading to a list of alphabets $A = \{0, 1, 2, \dots, k-1\}$.
- Then, each $s_i \in A$.

ϵ -Machine – Reconstruction

$S = 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, \dots$

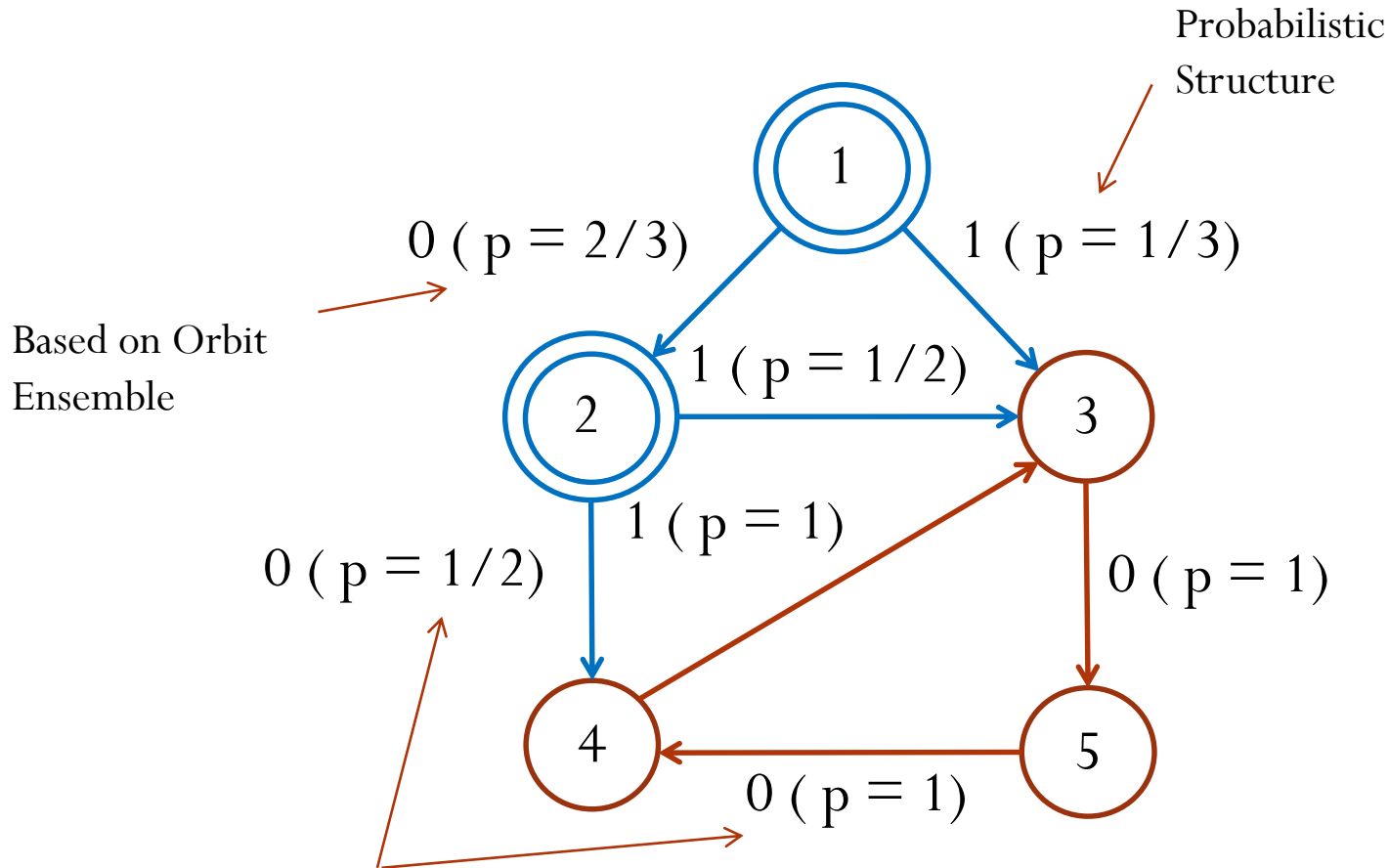
Period-3 sequence



- From symbol sequence \rightarrow build a *tree structure* with nodes and links .
- The links are given symbols according to the symbol sequence with $A = \{0, 1\}$.

\mathcal{E} -Machine – Reconstruction

$S = 0, 0, 1, 0, 0, 1, 0, 0, 1, 0, 0, 1, \dots$



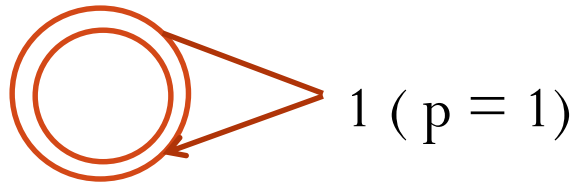
Combination of both deterministic and random computational resources

But the Probabilistic Structure is transient.

\mathcal{E} -Machine — Simplest Cases

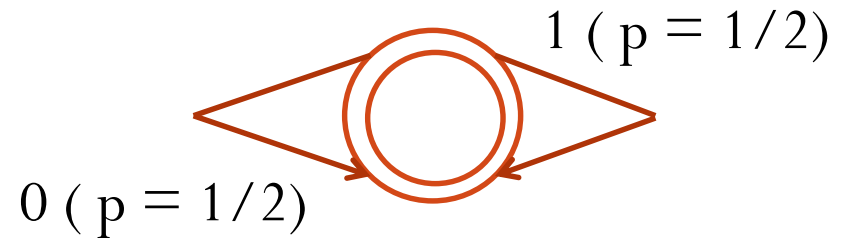
All Head or All “1” Process

$s = 1, 1, 1, 1, 1, 1, \dots$



Purely Random Process

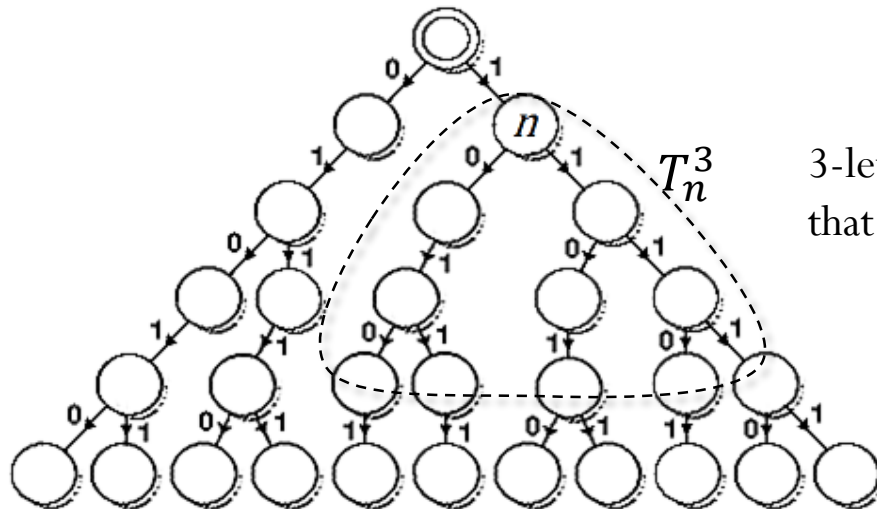
Fair Coin Process



- These are the extreme cases of Complexity.
- They are structurally simple.
- We expect their Complexity Measure to be zero.

\mathcal{E} -Machine: Complexity from Deterministic Dynamics

- The Symbol Sequence is derived from the Logistic Map.
- The parameter r is set to the band merging regime $r = 3.67859 \dots$
- For a finite sequence, a tree length $Tlen$ and a machine length $Mlen$ need to be defined.

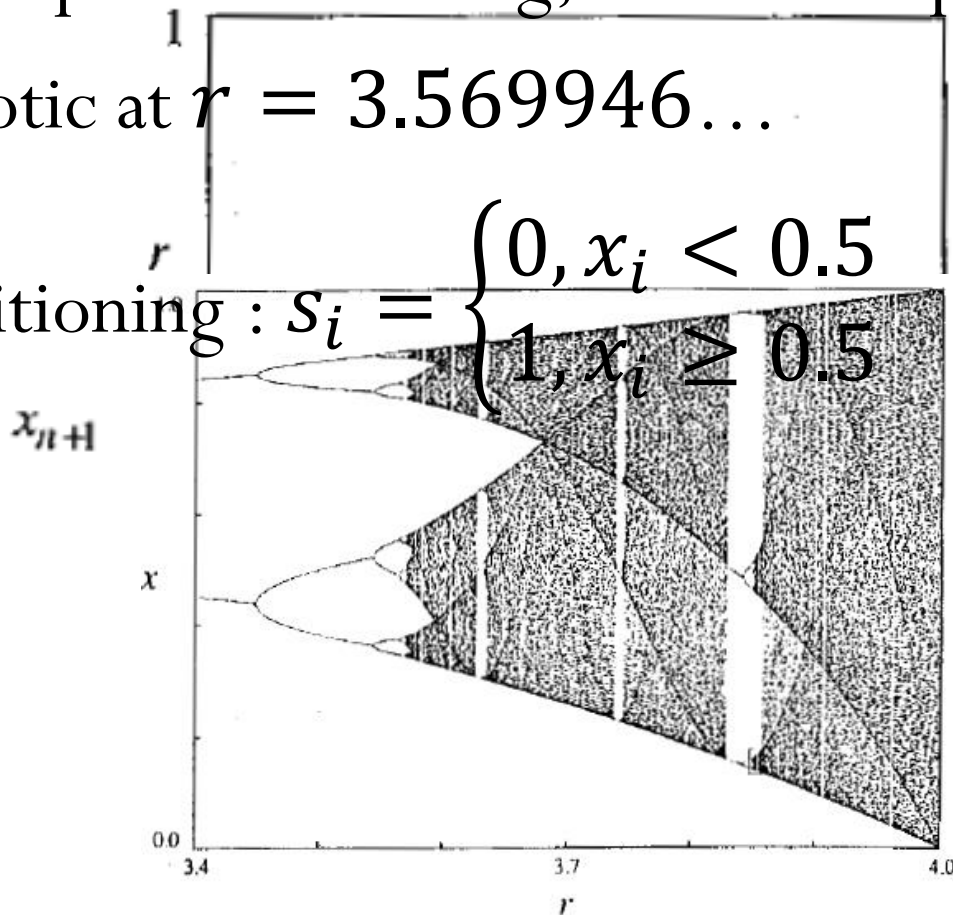


3-level sub-tree of n contains all nodes that can be reached within 3 links

The Logistic Map

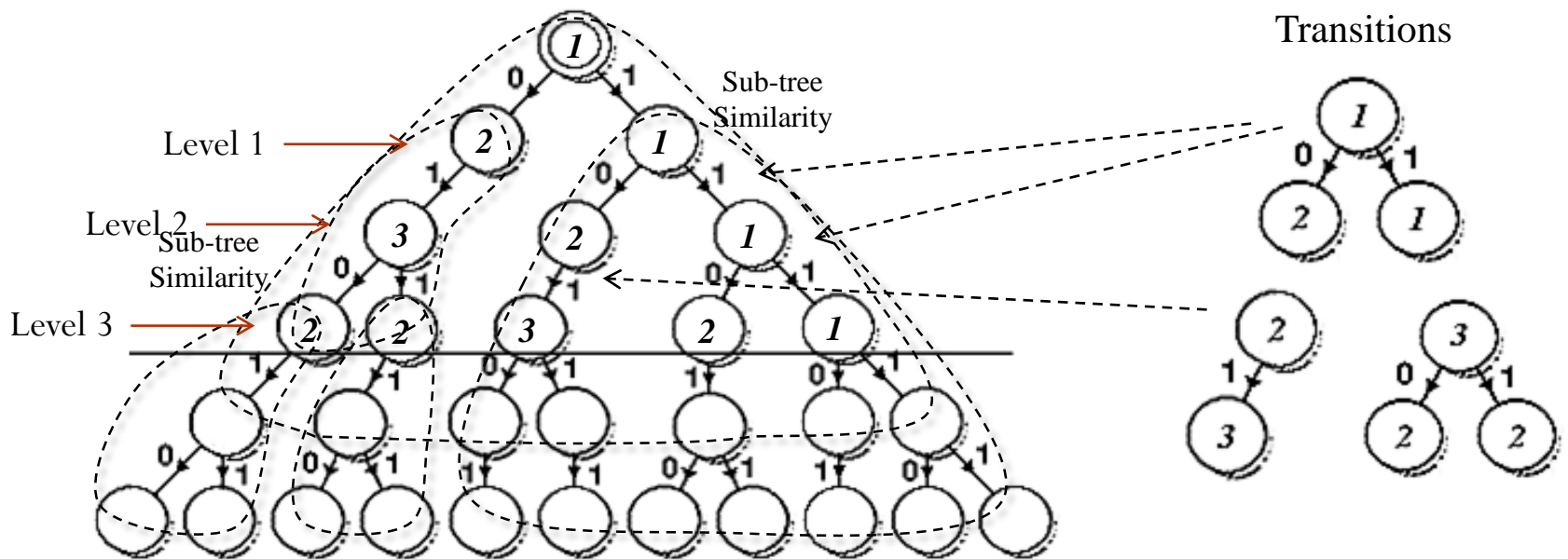
- Logistic equation : $x_{n+1} = rx_n(1 - x_n)$
- Starts period-doubling; at $r = 3$: period-2
- Chaotic at $r = 3.569946\dots$

- Partitioning : $S_i = \begin{cases} 0, & x_i < 0.5 \\ 1, & x_i \geq 0.5 \end{cases}$

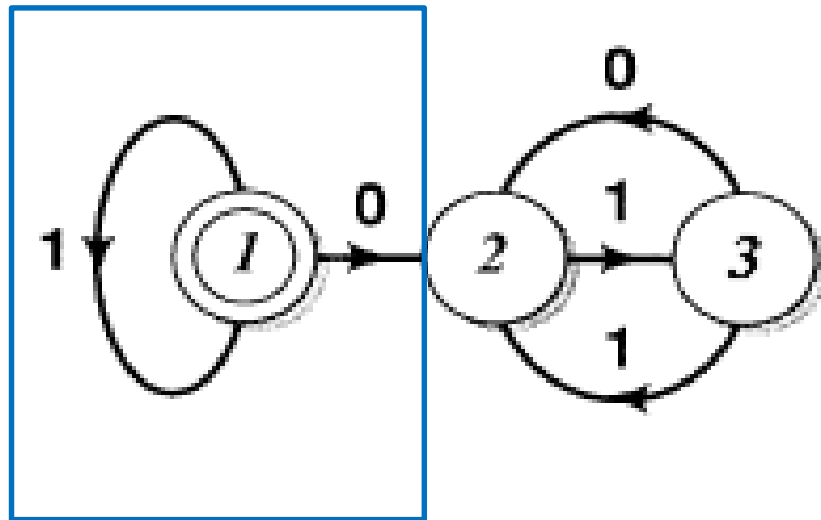


ϵ -Machine: Complexity from Deterministic Dynamics

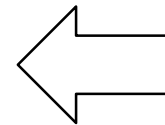
Construction of ϵ -machine:



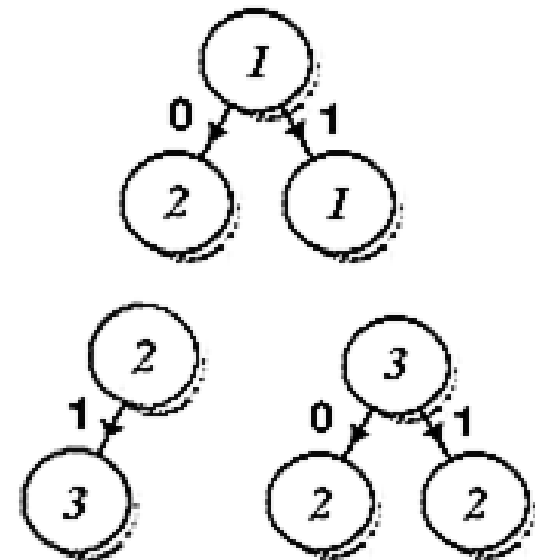
ϵ -Machine: Complexity from Deterministic Dynamics



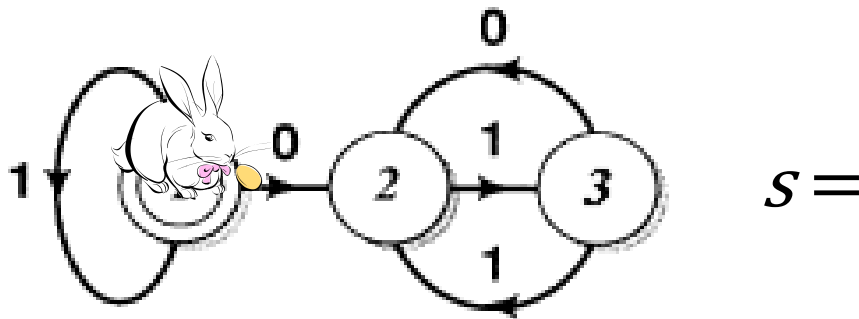
Transient State



Transitions



\mathcal{E} -Machine: Complexity from Deterministic Dynamics



- The \mathcal{E} -machine captures the patterns of the process.

\mathcal{E} -Machine: Causal State Splitting Reconstruction

Total = 9996

$$P(0|*) = 3309/9996 \approx 0.33$$

$$P(1|*) = 6687/9996 \approx 0.67$$

$$P(0|*0) = 1654/3309 \approx 0.5$$

$$P(1|*0) = 1655/3309 \approx 0.5$$

$$P(1|*1) = 5032/6687 \approx 0.75$$

$$P(0|*1) = 1655/6687 \approx 0.25$$

$$P(0|*00) \approx 0.51$$

$$P(1|*00) \approx 0.49$$

$$P(0|*01) = 0$$

$$P(1|*01) = 1$$

$$P(1|*10) \approx 0.51$$

$$P(0|*10) \approx 0.49$$

$$P(1|*11) \approx 0.67$$

$$P(0|*11) \approx 0.33$$

$$P(0|*000) \approx 0.49$$

$$P(0|*001) \approx 0$$

$$P(0|*010) \approx 0$$

$$P(0|*011) \approx 0.49$$

$$P(1|*100) \approx 0.49$$

$$P(0|*101) \approx 0$$

$$P(1|*110) \approx 0.51$$

$$P(1|*111) \approx 0.75$$

$$P(1|*000) \approx 0.51$$

$$P(1|*001) \approx 1$$

$$P(1|*010) \approx 0$$

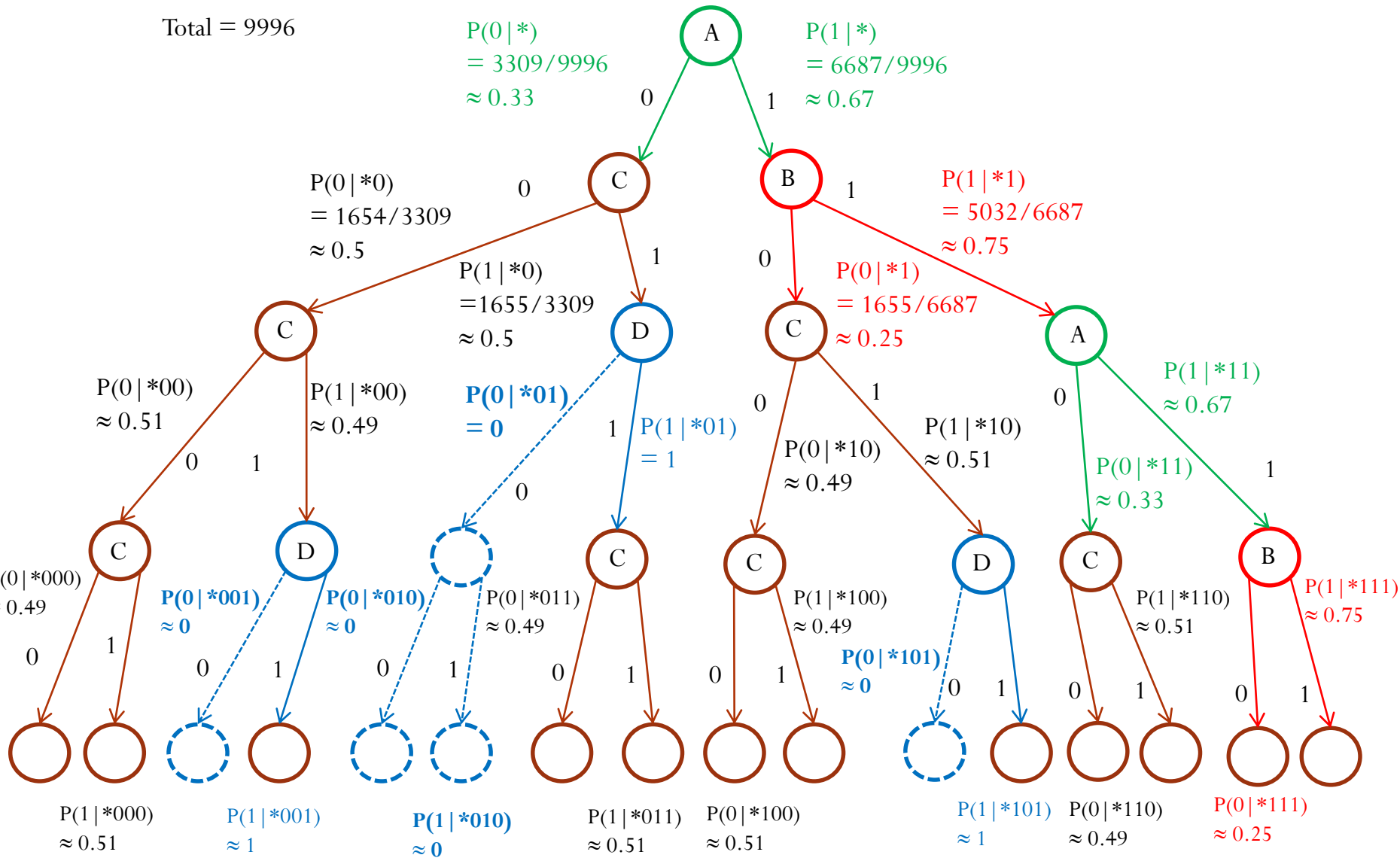
$$P(1|*011) \approx 0.51$$

$$P(0|*100) \approx 0.51$$

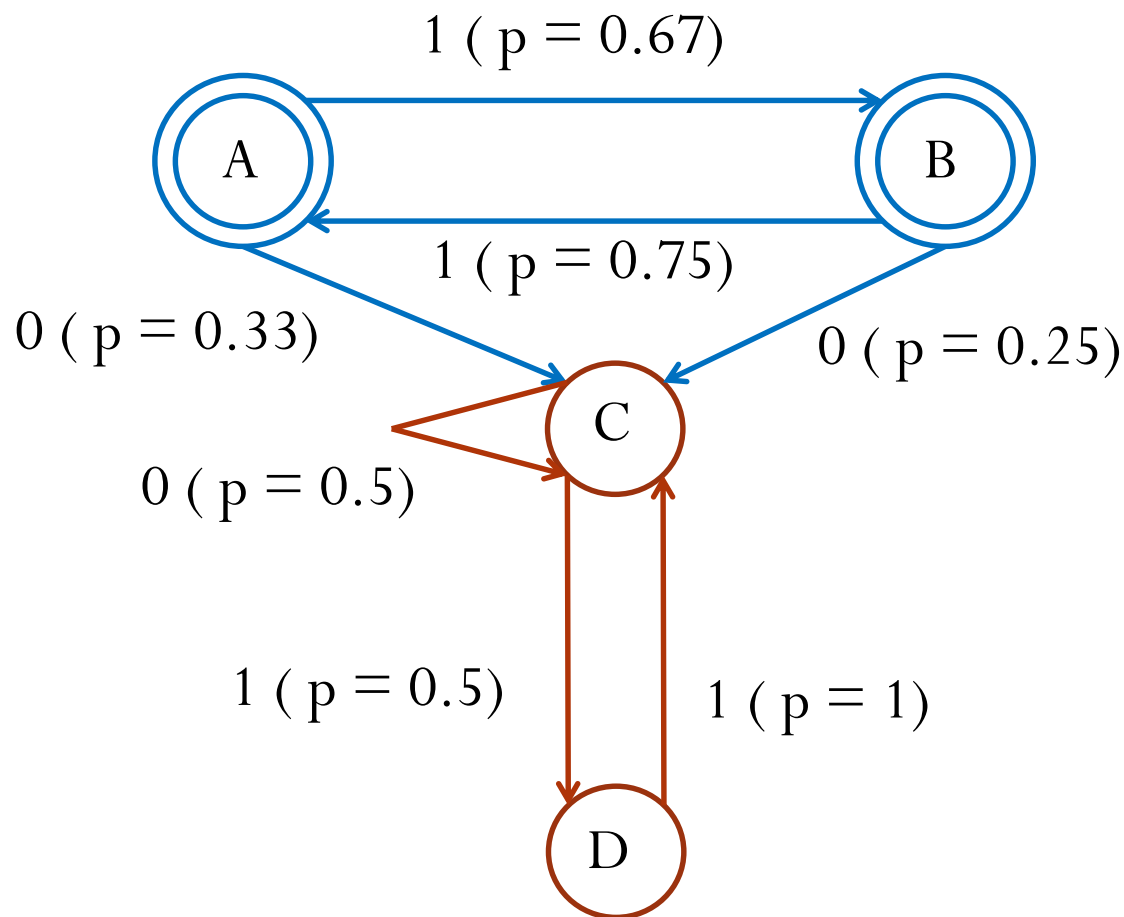
$$P(1|*101) \approx 1$$

$$P(0|*110) \approx 0.49$$

$$P(0|*111) \approx 0.25$$



\mathcal{E} -Machine: Causal State Splitting Reconstruction - The Even Process



Statistical Complexity and Metric Entropy

- Statistical Complexity is defined as

$$C_\mu = -\sum_{\{\sigma\}} p(\sigma) \log_2 p(\sigma) \quad \text{Bits}$$

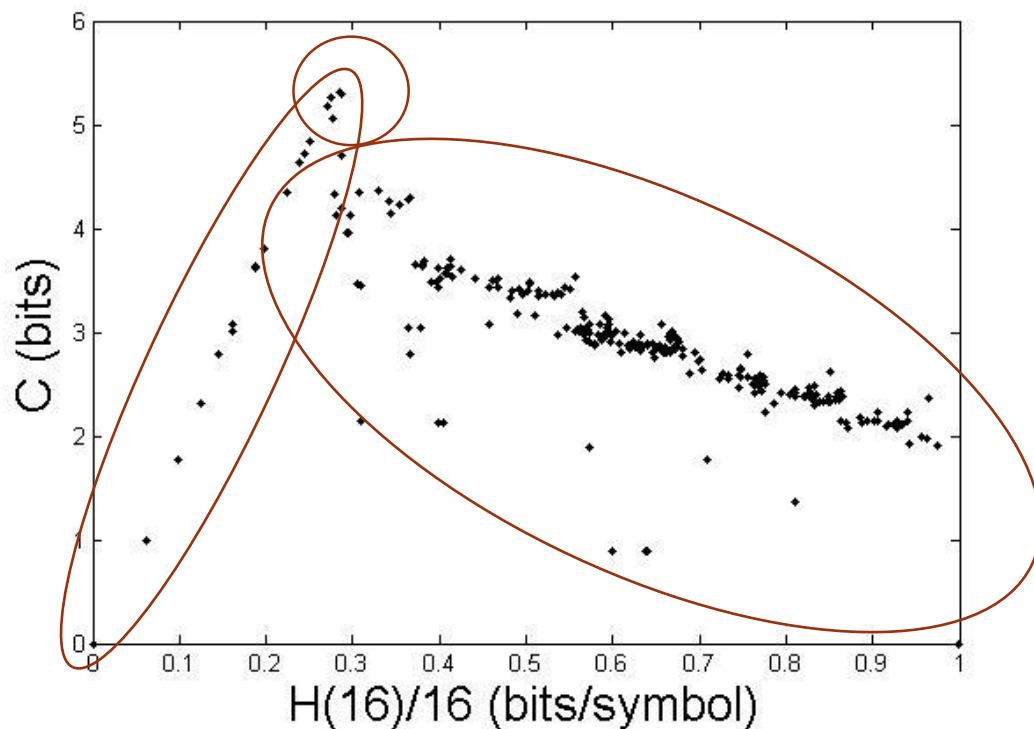
It serves to measure the computational resources required to reproduce a data stream.

- Metric Entropy is defined as

$$h = -\frac{1}{m} \sum_{\{s^m\}} p(s^m) \log_2 p(s^m) \quad \text{Bits/Symbols}$$

It serves to measure the diversity of observed patterns in a data stream.

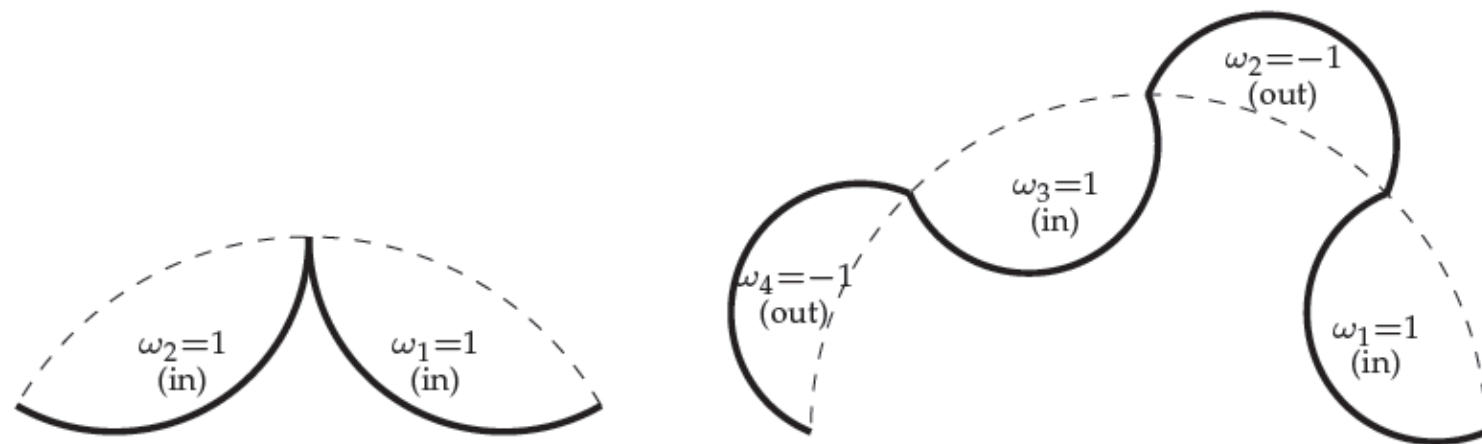
The Logistic Map



- Linear region in the periodic window
- Decreasing trend at chaotic region
- Phase transition, $H^* \approx 0.28$: “edge of chaos”

The Arc-Fractal Systems

A novel idea to generate fractals with “tunable” dimension¹



(a) $\alpha = \frac{2\pi}{3}$, $n = 2$ and $\omega = (\omega_1, \omega_2) = (1, 1)$

(b) $\alpha = \pi$, $n = 4$ and $\omega = (\omega_1, \omega_2, \omega_3, \omega_4) = (1, -1, 1, -1)$

Figure : Basic idea of the arc-fractal system.

- The parameters: α (angle), n (number of arcs) and ω (orientation of arcs)
- Three types of rule: single, multiple and random

¹ H. N. Huynh and L. Y. Chew, *Fractals*, **19** 141 (2011).

The Arc-Fractal Systems

Different parameters yield different fractals¹

Table : The combinations of α , n , and ω that give the classic fractals.

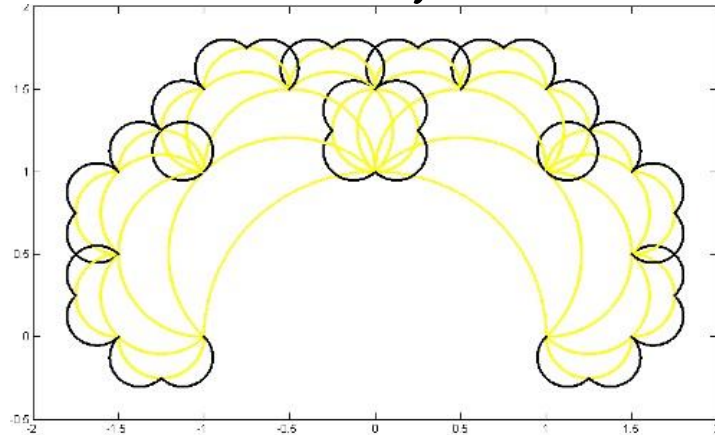
Classic fractals	α	n	ω
Koch	$\frac{2\pi}{3}$	2	(+1; +1)
Heighway	π	2	(-1; +1)
Lévy	π	2	(-1; -1)
Sierpiński	π	3	(+1; -1; +1)
Eisenstein*	π	3	(-1; +1; -1)

*Eisenstein fraction is created by joining three pieces of crab fractals together.

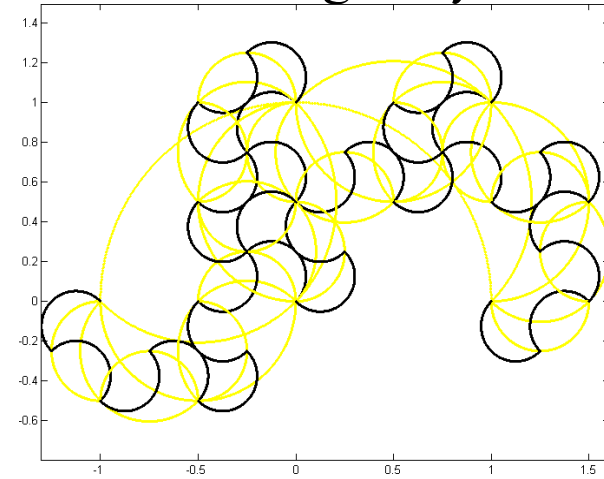
¹ H. N. Huynh and L. Y. Chew, *Fractals*, **19** 141 (2011).

The Arc-Fractal Systems

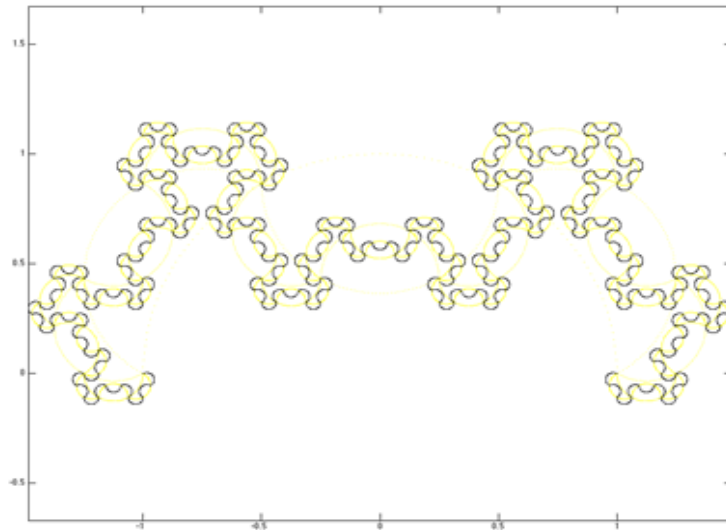
- Out-Out (Levy):



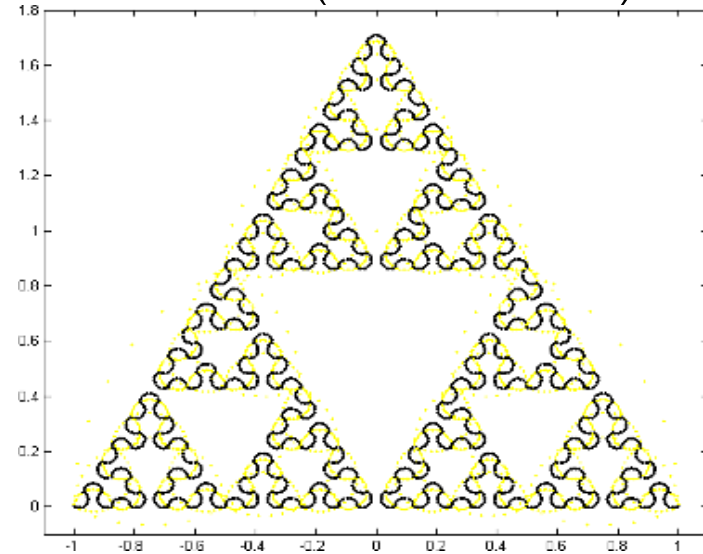
- Out-In (Heighway):



- Out-In-Out (Crab):

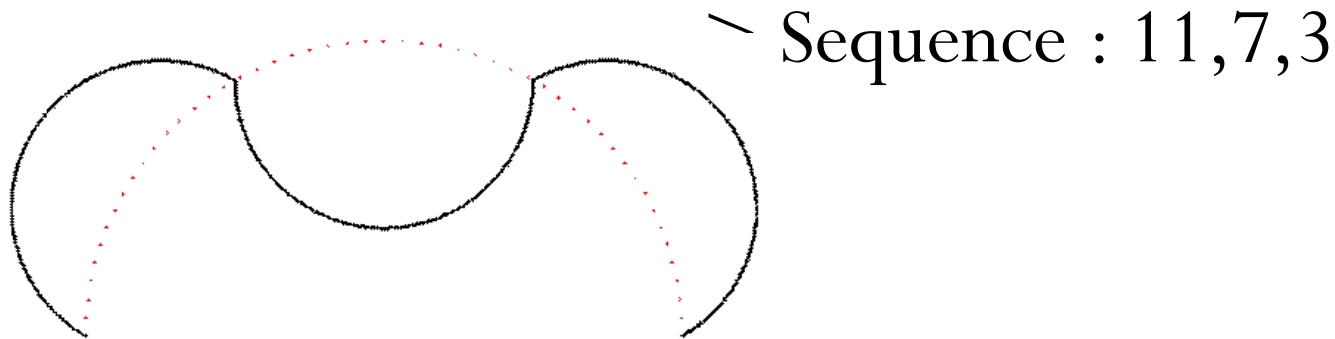


- In-Out-In (Arrowhead):



The Arc-Fractal Systems

- Sequence \rightarrow each arc is given an index according to its angle of orientation



Out-In-Out rule, level 3 of construction:

(7,3,11,7,11,3,11,7,3,11,3,7,11,7,3,7,11,3,11,7,3,11,3,7,3,11,7)

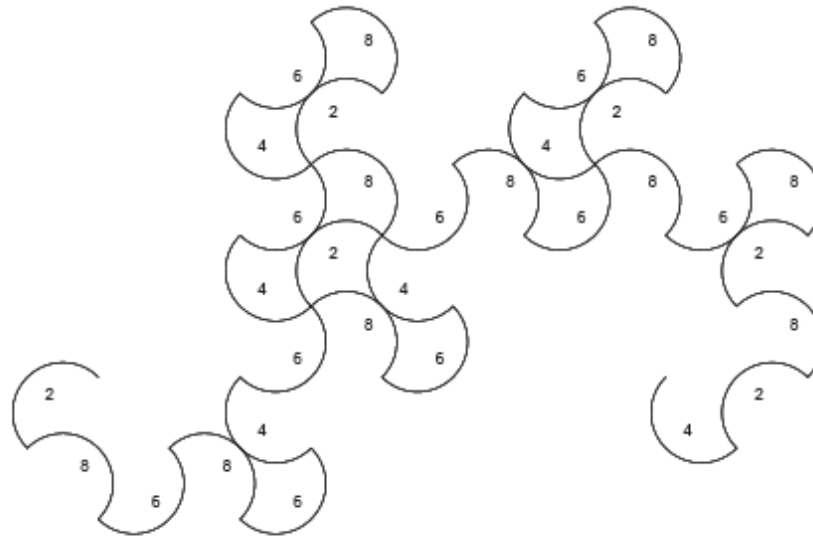
In base-3:

(1,0,2,1,2,0,2,1,0,2,0,1,2,1,0,1,2,0,2,1,0,2,0,1,0,2,1)

The Arc-Fractal Systems

- The fractals generated by the arc-fractal systems can be associated with symbolic sequences^{1,2}.

42828682864686828646424686468682



¹H. N. Huynh and L.Y. Chew, *Fractals* **19**, 141 (2011).

²H. N. Huynh, A. Pradana and L.Y. Chew, *PLoS ONE* **10**(2), e0117365 (2015)

The Arc-Fractal Systems

Table : Results of arc-fractal sequences with 2 arc segments (Heighway and Lévy) and 3 arc segments (arrowhead and crab).¹

Fractal	d	n	ω	number of symbols [†]	h (bit/symbol)	C_S (bit)
Lévy	2	2	(-1; -1)	4	0.389	9.80
Heighway	2	2	(-1; 1)	4	0.359	9.27
Arrowhead	$\frac{\ln 3}{\ln 2}$	3	(1; -1; 1)	3	0.387	7.34
Crab	$\frac{\ln 2}{\ln 3}$	3	(-1; 1; -1)	3	0.387	7.34

Complex Symbolic Sequence and Neuro-behavior

The complex symbolic sequences can be employed as stimuli in neurobehavioural or human pattern recognition experiments.¹

- Map each symbol to a corresponding physical state
- Whether and how the human brain can perceive or predict the states

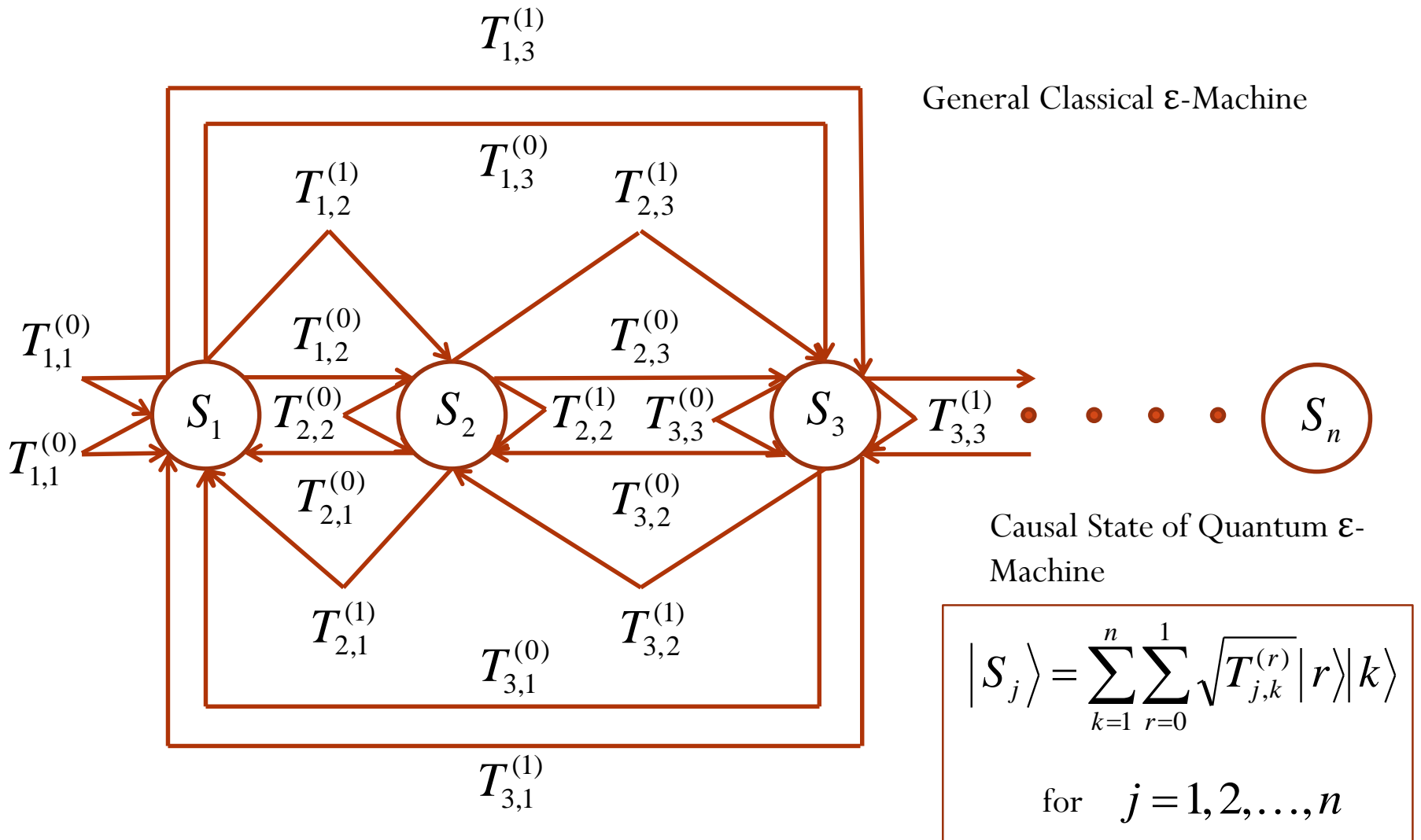
What is the next symbol in the sequence

◇♣♠♣♠♥♠♣♠♥◇♥♠♥♠♣♠♥◇♥◇ ...

- a. ♣ b. ◇ c. ♥ d. ♠? How confident are you?

Difficulty quantified by measure of complexity?

Quantum \mathcal{E} -Machine



Quantum Complexity

Quantum Causal State :

$$|S_j\rangle = \sum_{k=1}^n \sum_{r=0}^1 \sqrt{T_{j,k}^{(r)}} |r\rangle |k\rangle \quad \text{for } j = 1, 2, \dots, n$$

Quantum Complexity :

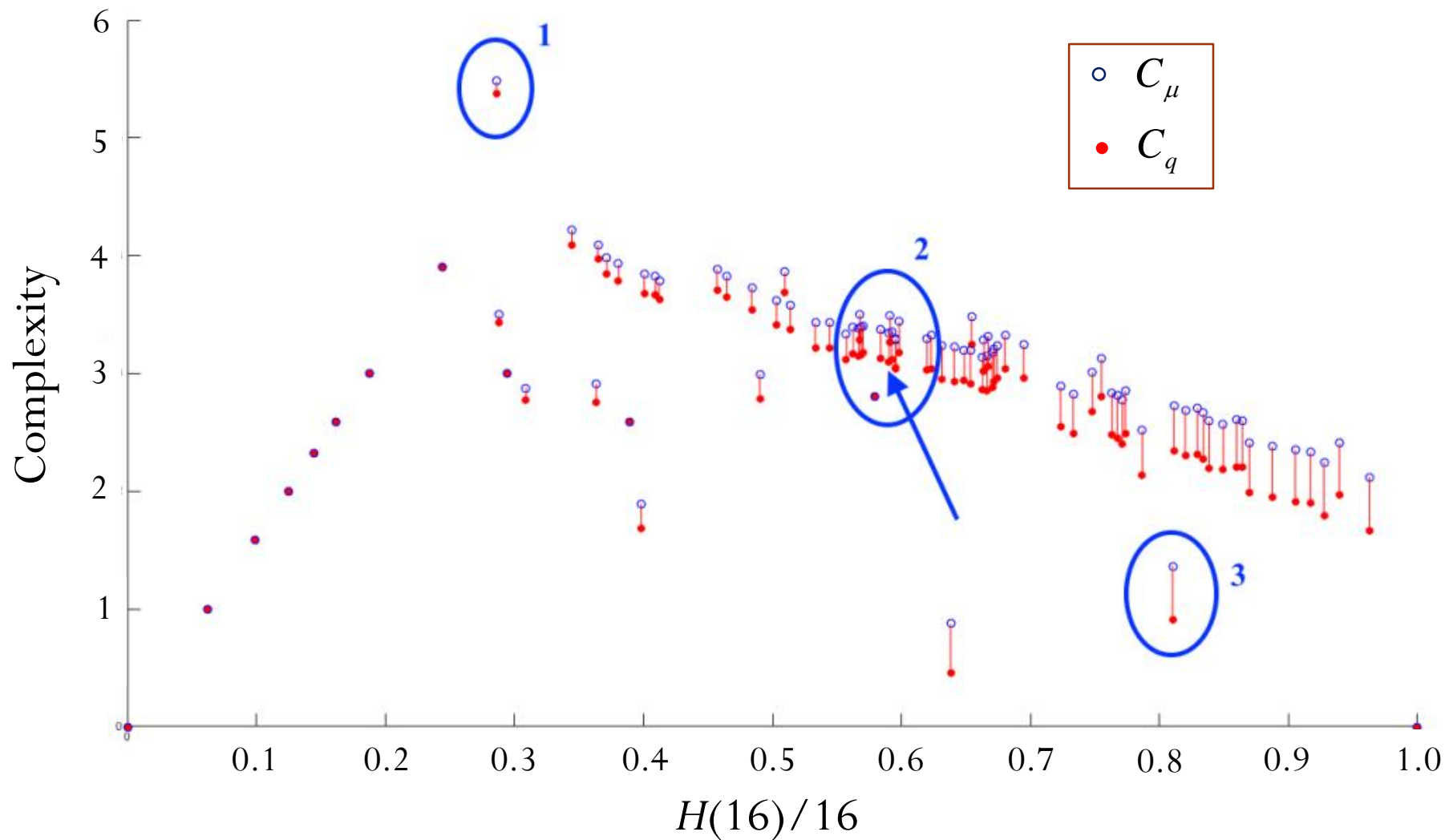
$$C_q = -\text{Tr}(\rho \log \rho) \quad \text{Bits}$$

where the density matrix $\rho = \sum_i p_i |S_i\rangle\langle S_i|$ and p_i is the probability of the quantum causal state $|S_i\rangle$.

Note that $C_q \leq C_\mu$ where $C_\mu = -\sum_{\{\sigma\}} p(\sigma) \log_2 p(\sigma)$ Bits

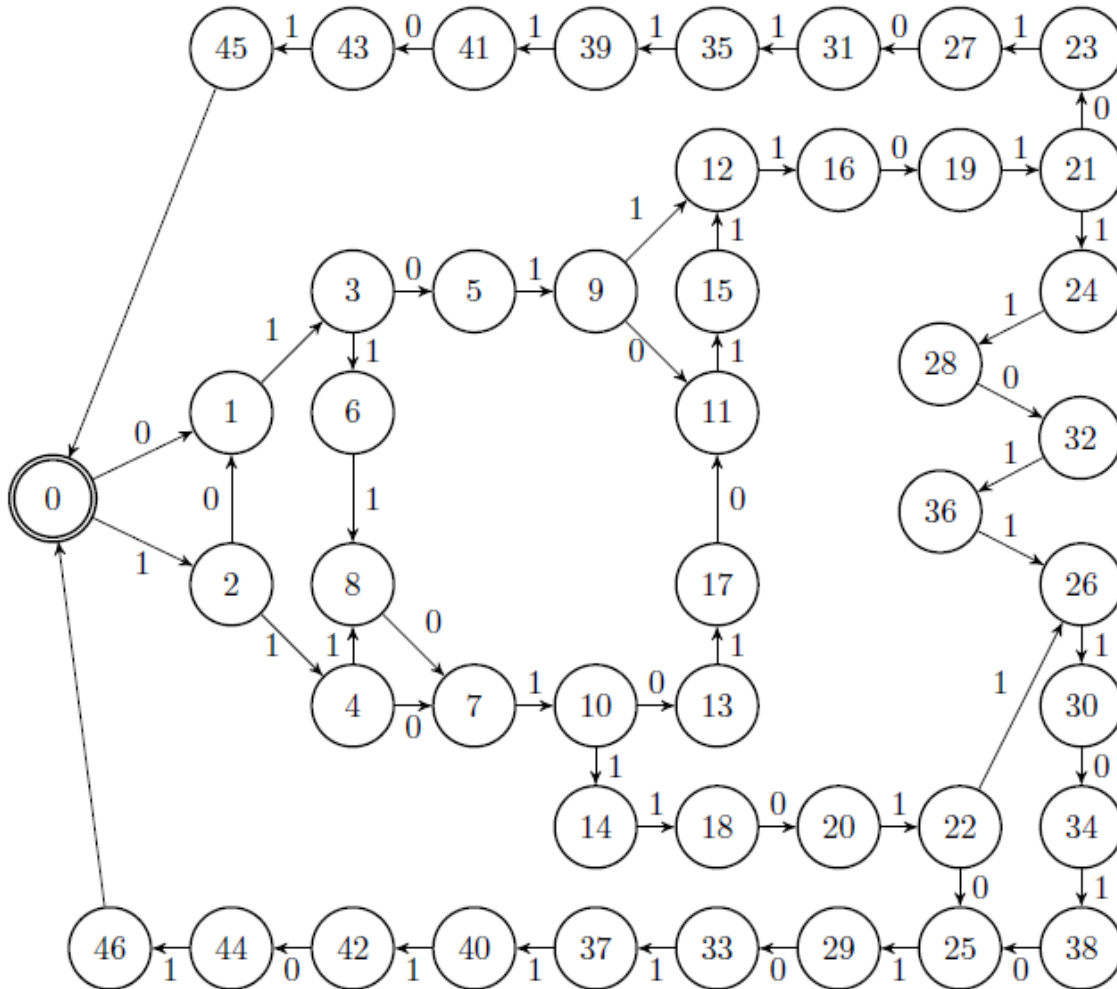
Quantum versus Classical Complexity

The Logistic Map



\mathcal{E} -Machine at Point 1

– Edge of Chaos

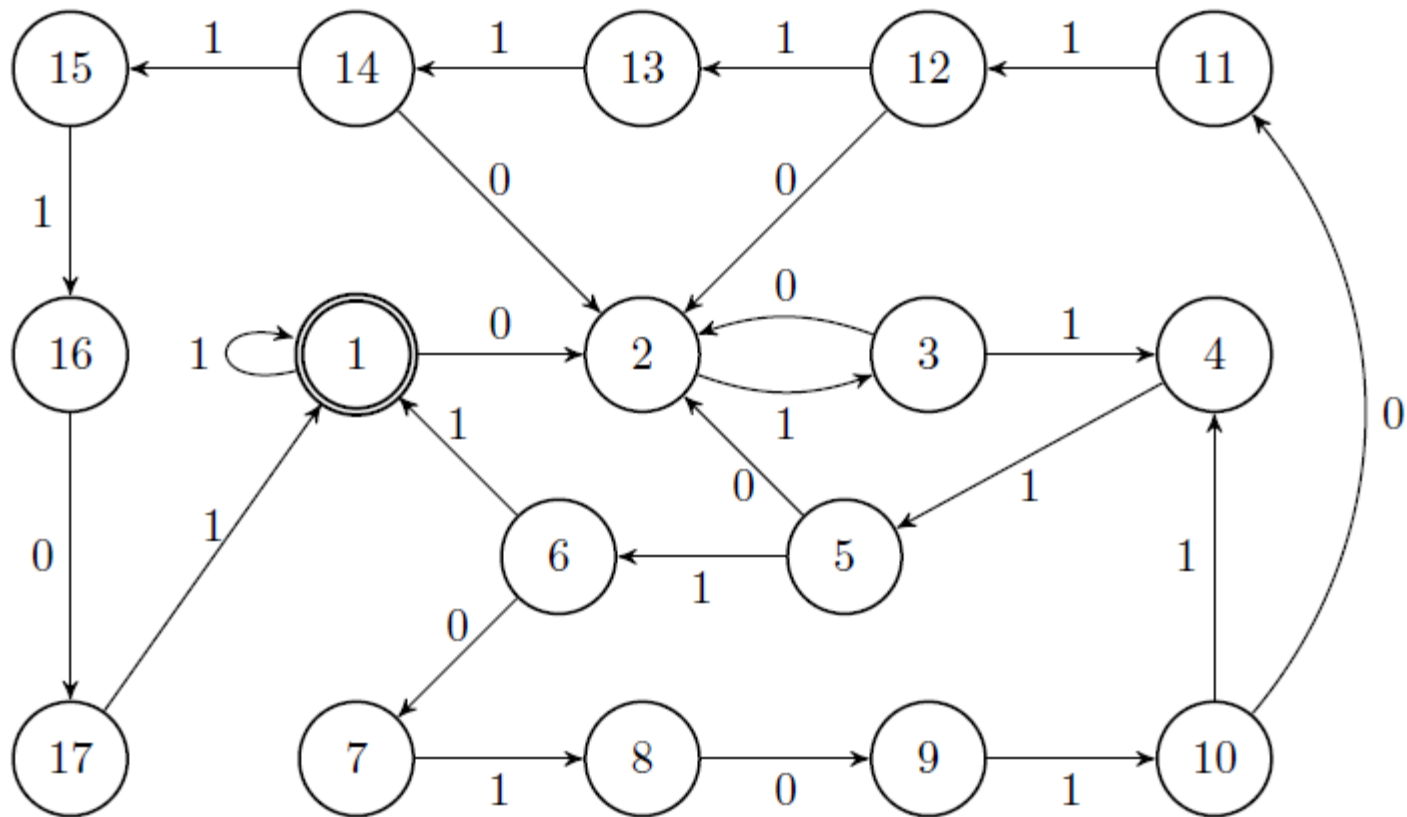


$$R = 3.569946\dots$$

$$C_\mu = 5.480236$$

$$C_q = 5.376835$$

ϵ -Machine at Point 2 – Chaotic Region



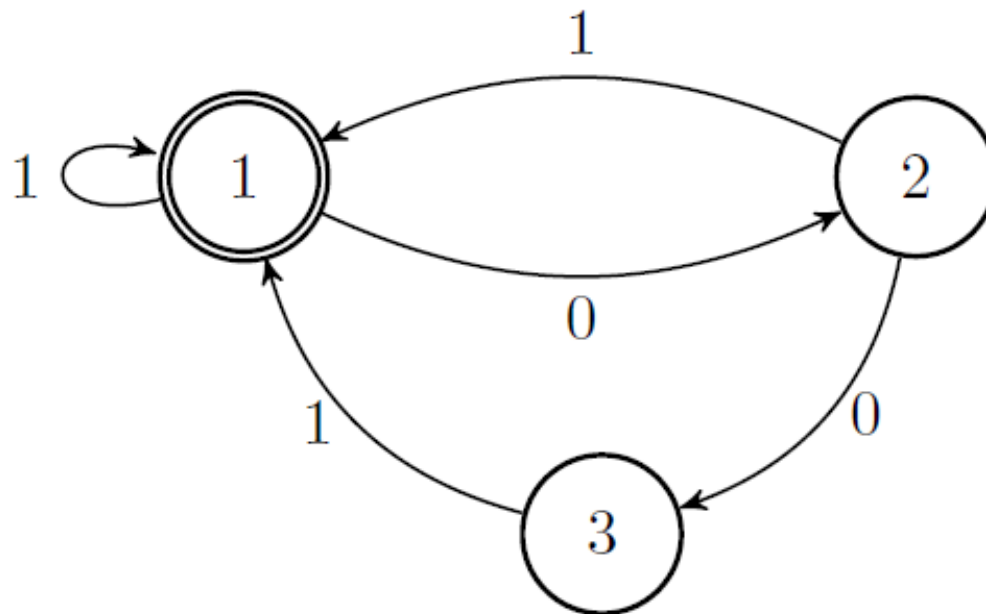
$$R = 3.715$$

$$C_{\mu} = 3.341681$$

$$C_q = 3.093695$$

\mathcal{E} -Machine at Point 3 – Chaotic Region

$$R = 3.960$$



$$C_{\mu} = 1.364846$$

$$C_q = 0.906989$$

Quantum \mathcal{E} -Machine

-Change of Measurement Basis

General Qubit Measurement Basis:

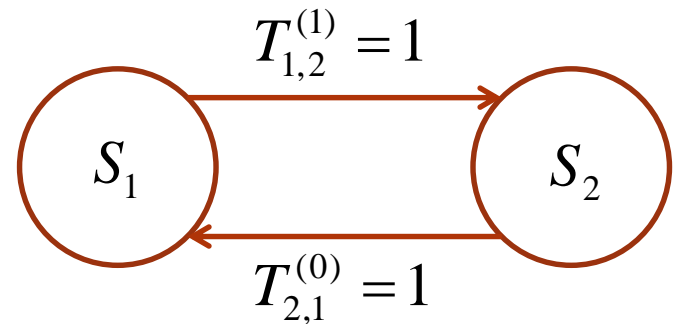
$$|\alpha\rangle = a|0\rangle + b|1\rangle$$

$$|\beta\rangle = b^*|0\rangle - a^*|1\rangle$$

where $|\alpha\rangle$ and $|\beta\rangle$ are new orthogonal measurement basis such that:

$$|0\rangle = a^*|\alpha\rangle + b|\beta\rangle$$

$$|1\rangle = b^*|\alpha\rangle - a|\beta\rangle$$



$$|S_1\rangle = |1\rangle|2\rangle$$

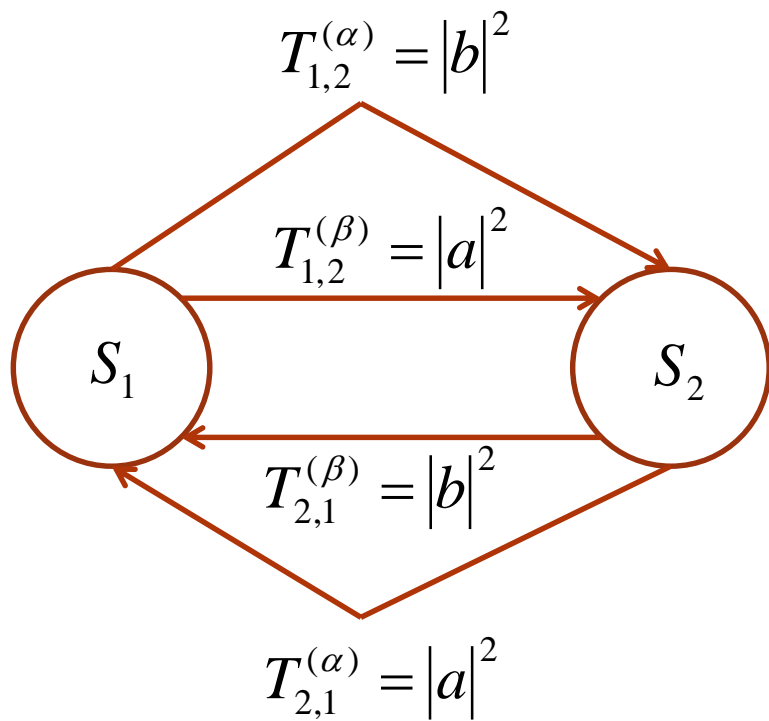
$$|S_2\rangle = |0\rangle|1\rangle$$

Quantum \mathcal{E} -Machine

-Change of Measurement Basis

$$|S_1\rangle = b^*|\alpha\rangle|2\rangle - a|\beta\rangle|2\rangle$$

$$|S_2\rangle = a^*|\alpha\rangle|1\rangle + b|\beta\rangle|1\rangle$$



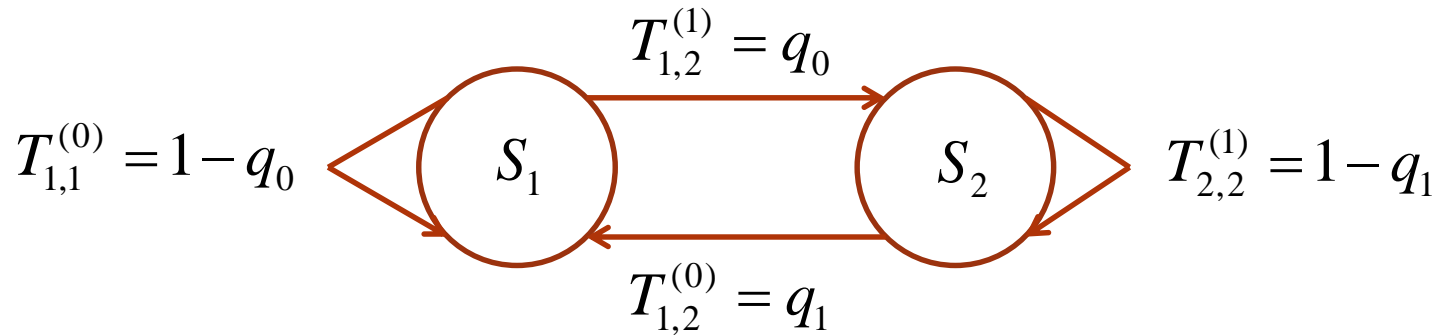
Deterministic \mathcal{E} - Machine



Stochastic \mathcal{E} - Machine

Quantum \mathcal{E} -Machine

-Change of Measurement Basis



Stochastic \mathcal{E} - Machine

$$|S_1\rangle = (1 - q_0)|0\rangle|1\rangle + q_0|1\rangle|2\rangle$$

$$|S_2\rangle = q_1|0\rangle|1\rangle + (1 - q_1)|1\rangle|2\rangle$$

Quantum \mathcal{E} -Machine

-Change of Measurement Basis

$$|S_1\rangle = \{(1 - q_0)a^*|1\rangle + q_0 b^*|2\rangle\}|\alpha\rangle + \{(1 - q_0)b|1\rangle - q_0 a|2\rangle\}|\beta\rangle$$

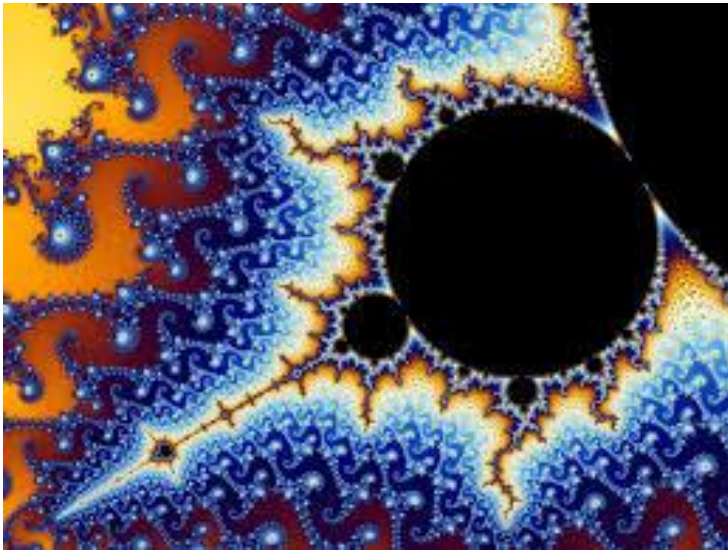
$$|S_2\rangle = \{q_1 a^*|1\rangle + (1 - q_1)b^*|2\rangle\}|\alpha\rangle + \{q_1 b|1\rangle - (1 - q_1)a|2\rangle\}|\beta\rangle$$

If the quantum \mathcal{E} – machine is prepared in the quantum causal state $|S_1\rangle$ and measurement yields $|\alpha\rangle$, the quantum \mathcal{E} – machine now possesses the causal state:

$$(1 - q_0)a^*|S_1\rangle + q_0 b^*|S_2\rangle$$

which is a superposition of the quantum resource. Further measurements in the $|\alpha\rangle$ and $|\beta\rangle$ basis leads to further iterations within the above results due to the inherent entangling features in the quantum causal state structure.

Ideas to Proceed



Mandelbrot Set

- *Complexity within Simplicity*

Question:

Could there be infinite regress in terms of information processing within quantum ϵ – Machine?

Question:

Could classical ϵ – Machine acts as a mind-reading machine in the sense of Claude Shannon's? Would a quantum version do better?





Neil Huynh



Andri Pradana



Matthew Ho

Thank You