

Model of Computation

- i) read input from left to right
- ii) has a working tape
- iii) write output from left to right

Universal Computer U

• for each computer A , there is a program S_A satisfying

$$U: \exists A \cdot p \mapsto f_A(p)$$

For a computer A , it defines a func.

$$f_A: \{0,1\}^* \rightarrow \{0,1\}^* \cup \{0,1\}^\infty$$

Def) The Kolmogorov complexity $K_U(x)$ of a string x w.r.t. a uni. comp. U is

$$K_U(x) := \min_{p: U(p)=x} l(p)$$

Def) The Conditional \sim

$$K_U(x|d(x)) := \min_{p: U(p,d(x))=x} l(p)$$

(my understand: there is some mapping $(p,d) \mapsto pe$)

Thm) (Universality of K_U)

U is univ. $\rightarrow \forall A$: computer. $\exists c_A \cdot \forall x \cdot K_U(x) \leq K_A(x) + c_A$

Corollary) U, A are univ. $\rightarrow \exists c \cdot \forall x \cdot |K_U(x) - K_A(x)| < c$

pf) $f_A(p_A) = x, l(p_A) = K_A(x)$

$$l(S_A \cdot p_A) = l(S_A) + l(p_A) = c_A + l(p_A)$$

$$K_U(x) = \min_{p: U(p)=x} l(p) \leq l(S_A \cdot p_A) = c_A + l(p_A) = c_A + K_A(x)$$

Thm. $K(x|d(x)) \leq l(x) + c$

Thm. (Ubd of K)

pf. Print x

$$K(x) \leq K(x|d(x)) + \log^* l(x) + c$$

pf. desc. $l(x)$ as "... [log log log] [log l(x) bits]"

Thm. (Lbd of K) $|\{x \in \{0,1\}^* \mid K(x) < k\}| < 2^k$

$$pf) |\{x \mid K(x) < k\}| \leq |\{p \mid l(p) < k\}| < 2^k$$

Notation: $H_0(p) := -p \log p - (1-p) \log (1-p)$

That is, $H_0(x)$ for a rand. var. X , is a rand. var.

Fact:

$$\sqrt{\frac{n}{8k(n-k)}} \cdot 2^{nH(k/n)} \leq \binom{n}{k} \leq \sqrt{\frac{n}{2k(n-k)}} \cdot 2^{nH(k/n)}$$

Example (Seq. of n -bits w/ k ones) $K(x^{(n)}) \leq n H_0(\frac{k}{n}) + \frac{1}{2} \log n$

p : Print the i -th of seq. of n -bits w/ k ones

$$l(p) = c + \log n + \log \binom{n}{k} \quad (n \text{ is known})$$

do express k to expr. i

$$\leq c' + \log n + nH(k/n) - \frac{1}{2} \log n$$

$$= c' + \frac{1}{2} \log n + nH(k/n)$$

Thm $K(x^{(n)}|n) \leq n H_0(\frac{k}{n}) + \frac{1}{2} \log n + c$

$$\log \left(\sqrt{\frac{n}{2k(n-k)}} \cdot 2^{nH(k/n)} \right)$$

$$= \log \left(\frac{1}{\sqrt{2npg}} \cdot 2^{nH(k/n)} \right) \quad (p = \frac{k}{n}, q = 1-p)$$

$$= -\frac{1}{2} \log n - \frac{1}{2} \log pg + nH(k/n) + c_0$$

$$\leq -\frac{1}{2} \log n + nH(k/n) + c_0'$$

Lem. $\forall u$: computer, $\sum_{p:|u(p)| \leq n} 2^{-l(p)} \leq 1$ $\Rightarrow \{p:|u(p)| \leq n\}$ is prefix-free

Thm (Rel. of K and H)

$\{X_i\}$, $X_i \sim f(x)$, i.i.d. $x \in \mathcal{X}$, $|\mathcal{X}| < \infty$.

Let $f(x^{(n)}) = \prod f(x_i)$

$$\exists C. H(X) \leq \frac{1}{n} \sum_{x^{(n)}} f(x^{(n)}) K(x^{(n)}|n) \leq H(X) + \frac{(|\mathcal{X}|-1) \log n}{n} + \frac{C}{n}$$

That is,

$$E\left(\frac{1}{n} \cdot K(x^{(n)}|n)\right) \rightarrow H(X)$$

pf) $H(X) \leq \frac{1}{n} \sum_{x^{(n)}} f(x^{(n)}) K(x^{(n)}|n)$

$$\sum_{x^{(n)}} f(x^{(n)}) K(x^{(n)}|n) = \sum_{x^{(n)}} f(x^{(n)}) \cdot |C(x^{(n)})| = E|C(X_1, \dots, X_n)| \leq H(X_1, \dots, X_n) = nH(X)$$

$C: \mathcal{X}^{(n)} \mapsto \mathcal{P} \in \{0, 1\}^*$, prefix-free

Theory of source coding

$$\left(\frac{1}{n} \sum_{x^{(n)}} f(x^{(n)}) K(x^{(n)}|n) \leq H(X) + \frac{(|\mathcal{X}|-1) \log n}{n} + \frac{C}{n}\right)$$

Sps X_i is bin.

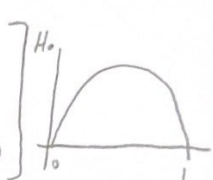
$$K(x^{(n)}|n) \leq nH_0\left(\frac{1}{n} \sum x_i\right) + \frac{1}{2} \log n + C$$

$$EK(X_1, \dots, X_n|n) \leq nEH_0\left(\frac{1}{n} \sum X_i\right) + \frac{1}{2} \log n + C$$

$$\leq nH_0\left(\frac{1}{n} \sum EX_i\right) + \frac{1}{2} \log n + C$$

$$= nH_0(\theta) + \frac{1}{2} \log n + C \quad (X_1, \dots, X_n \text{ are i.i.d. } \sim \text{Bernoulli}(\theta))$$

Jensen's inequ.
 $E(\varphi(x)) \leq \varphi(E(x))$ when φ is concave
 Concavity of H_0
 $H_0(\lambda p_1 + (1-\lambda)p_2) \geq \lambda H_0(p_1) + (1-\lambda)H_0(p_2)$



If \mathcal{X} is not bin.

$$K(x^{(n)}|n) \leq nH(P_{\mathcal{X}^{(n)}}) + \frac{(|\mathcal{X}|-1) \log n}{n} + C$$

$$\left(\begin{array}{l} \frac{(\#i)-1}{\# \text{ of } i\text{-th symbol}} \\ \leq 2^{nH(P_{\mathcal{X}^{(n)}})} \end{array} \right)$$

\downarrow

$\rightarrow ?$

$$EK(x^{(n)}|n) \leq nH(X) + (|\mathcal{X}|-1) \log n + C$$

Corollary) $E\left(\frac{1}{n} K(n^{(n)})\right) \rightarrow H(X)$

Def. For integer n , $K(n) = \min_{p \in \mathcal{U}(n)} l(p)$

Thm. $K(n) \leq K_A(n) + C_A$, \mathcal{U}, A : univ.

Thm. $K(n) \leq \log^* n + C$

Thm. There are an inf # of ints n s.t. $K(n) > \log(n)$

pf) Note that $\sum_n 2^{-K(n)} \leq 1$.

Sps $K(n) < \log n$ for all $n > n_0$, then

$$\sum_{n=n_0}^{\infty} 2^{-K(n)} > \sum_{n=n_0}^{\infty} 2^{-\log n} = \sum_{n=n_0}^{\infty} \frac{1}{n} = \infty \quad \text{contradiction}$$

Thm) $X_1, \dots, X_n \sim \text{Bernoulli}(\frac{1}{2})$

$$P(K(X_1, X_2, \dots, X_n | n) < n-k) < 2^{-k}$$

pf) $P(K(X_1, \dots, X_n | n) < n-k)$

$$= \sum_{x^{(n)}: K(x^{(n)}) < n-k} 2^{-n}$$

$$= |\{x^{(n)} \mid K(x^{(n)}) < n-k\}| \cdot 2^{-n}$$

$$< 2^{n-k} \cdot 2^{-n} = 2^{-k}$$

Def) $x^{(n)}$ is algorithmically random if $K(x^{(n)} | n) \geq n$

Def) $x \in \{0,1\}^\infty$ is incompressible if $\lim_{n \rightarrow \infty} \frac{K(x_1, \dots, x_n | n)}{n} = 1$

Thm) (Strong law of large #s for incompressible seq.)

x_1, x_2, \dots is incomp. implies $\frac{1}{n} \sum_{i=1}^n x_i \rightarrow \frac{1}{2}$

pf) Let $\theta_n = \frac{1}{n} \sum_{i=1}^n x_i$

* typo in the book: $\frac{2 \log n}{n}$

$$\frac{K(x^{(n)} | n)}{n} < H_0(\theta_n) + \frac{\log n}{2n} + \frac{c'}{n}$$

Since x_i is incomp.,

$$\forall \epsilon. \exists n. 1 - \epsilon \leq \frac{K(x^{(n)} | n)}{n} \leq H_0(\theta_n) + \frac{\log n}{2n} + \frac{c'}{n}$$

Thus

$$\forall \epsilon. \exists n. H_0(\theta_n) > 1 - \frac{\frac{1}{2} \log n + c'}{n} - \epsilon$$

$$\nRightarrow \forall \epsilon. \exists n. H_0(\theta_n) > 1 - \epsilon$$

$$\nRightarrow \forall \epsilon. \exists n. \theta_n \in (\frac{1}{2} - \epsilon, \frac{1}{2} + \epsilon)$$



Recall) $\{X_i\}$: i.i.d.

$$E \frac{1}{n} K(X^n | n) \rightarrow H(X)$$

They do not imply each other.

Thm) $\{X_i\}$ i.i.d. Bernoulli(θ)

$$\frac{1}{n} K(X^n | n) \rightarrow H_0(\theta) \text{ in probability}$$

pf) Let $\bar{X}_n = \frac{1}{n} \sum_{i=1}^n X_i$

* typo in the book: $\frac{1}{2} \log n$

$$K(X^n | n) \leq n H_0(\bar{X}_n) + \frac{1}{2} \log n + c$$

By weak law,

$$\bar{X}_n \rightarrow \theta \text{ in prob. } (\equiv \mathbb{P}(|\bar{X}_n - \theta| > \epsilon) = 1 \text{ for all } \epsilon > 0)$$

$$\text{Thus, } \mathbb{P}(K(X^n | n) \leq n H_0(\theta) + \frac{1}{2} \log n + c + \epsilon) = 1 \forall \epsilon > 0$$

$$\nRightarrow \mathbb{P}(K(X^n | n) \leq n H_0(\theta) + \epsilon) = 1 \forall \epsilon > 0$$

Now we have to show

$$\mathbb{P}(H_0(\theta) - \frac{1}{n} K(X^n | n) \leq \epsilon) = 1, \forall \epsilon > 0.$$

Let $A_\varepsilon^{(n)}$ be a typical set

$$A_\varepsilon^{(n)} = \{x^{(n)} \mid |-\frac{1}{n} \log p(x^{(n)}) - H(X)| < \varepsilon\}$$

Then
 i) $|A_\varepsilon^{(n)}| \geq (1-\varepsilon)2^{n(H(X)-\varepsilon)}$ ii) $\Pr[X^{(n)} \in A_\varepsilon^{(n)}] \geq 1-\varepsilon$

$$\Pr\{x^{(n)} \mid K(x^{(n)}|n) < n(H_0(\theta) - c)\} \leq 2^{-n(H_0(\theta) - c)}$$

For any fixed c , for any $\varepsilon > 0$, \exists large n

$$\begin{aligned} & \Pr(K(x^{(n)}|n) < n(H_0(\theta) - c)) \\ & \leq \Pr(X^{(n)} \notin A_\varepsilon^{(n)}) + \Pr(X^{(n)} \in A_\varepsilon^{(n)}, K(x^{(n)}|n) < n(H_0(\theta) - c)) \\ & \leq \varepsilon + \sum_{x^{(n)} \in A_\varepsilon^{(n)}, K(x^{(n)}|n) < n(H_0(\theta) - c)} p(x^{(n)}) \quad - \text{ii)} \\ & \leq \varepsilon + \sum_{\substack{x^{(n)} \in A_\varepsilon^{(n)} \\ K(x^{(n)}|n) < n(H_0(\theta) - c)}} 2^{-n(H_0(\theta) - \varepsilon)} \quad - \text{def} \\ & \leq \varepsilon + 2^{n(H_0(\theta) - c)} \cdot 2^{-n(H_0(\theta) - \varepsilon)} \quad - \text{iii)} \\ & = \varepsilon + 2^{-n(c - \varepsilon)} \end{aligned}$$

Final
 = ...
 Lbd of K

That is, $\Pr(K(x^{(n)}|n) < n(H_0(\theta) - c))$

$$= \Pr(c < H_0(\theta) - \frac{1}{n} K(x^{(n)}|n)) \rightarrow 0 \text{ as } n \rightarrow \infty \quad \square$$

Def) The universal probability of a string x (on an universal comp).

$$P_U(x) = \sum_{p: U(p)=x} 2^{-|p|} = \Pr[U(p)=x].$$

Thm For every computer A ,

$$P_U(x) \geq c_A P_A(x) \quad c_A \text{ dep. on } A \text{ and } U.$$

pf)
$$P_U(x) = \sum_{p: U(p)=x} 2^{-|p|} \geq \sum_{p: A(p)=x} 2^{-|p|-c_A} = c_A P_A(x).$$

↳ A 의 입력으로 U 가 입력될 때

Def. (Chaitin's #)

$$\Omega := \sum_{\substack{p: U(p) \\ \text{halts}}} 2^{-|p|} = \Pr[U(p) \text{ halts}],$$

Pr Bernoulli?

출력 방향 vs 입력 vs 프로그램 길이

- i) Ω is noncomputable
- ii) Ω is a "philosopher's stone"
- Ω_n of an n -bit string의 길이를 알 수 있다.
- (알 수 없는 길이를 by Gödel's incompleteness thm.)
- iii) Ω is algorithmically random.

Let me know Ω_n , where $|p|=n$.
 To know the program p halts,
 We run all programs p' with $|p'| \leq |p|$ parallelly.
 Note that we know how many p' 's are halts.
 We can run them until ~~halts~~
 with the number, the # of halts, p 's are the same

Thm $\exists c$ s.t $K(\Omega_n) > n - c. \forall n$

pf

Consider the following program:

let $a := w_1 \dots w_n$.
 find x_0 s.t. $K(x_0) > n$
 the smallest
 print x_0 .

def of x_0 .

This program has a complexity $K(\Omega_n) + c \geq K(x_0) > n$
 $\Rightarrow K(\Omega_n) > n - c.$