

# Algorithmic randomness and monotone complexity on product space

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## Contents:

1. Lambalgen's theorem for correlated probability
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Conditional probability:

$P$ : computable probability on  $X \times Y (= \Omega^2)$ .

marginal distributions

$$P_X(x) := P(x, \Omega).$$

$$P_Y(y) := P(\Omega, y).$$

conditional probability :

$$P(x|y) := P(x, y)/P_Y(y) \text{ for } y \in S$$

$$P(x|y^\infty) := \lim_{y \rightarrow y^\infty} P(x, y)/P_Y(y) \text{ for } y^\infty \in \Omega$$

(if the right-hand-side exists.)

Notation:

$$S := \{0, 1\}^*, \Omega := \{0, 1\}^\infty, x, y \in S, x^\infty, y^\infty \in \Omega.$$

$$\Delta(x) := \{xy^\infty | y^\infty \in \Omega\}, P(x, y) := P(\Delta(x), \Delta(y))$$

$\mathcal{R}^P$ : the set of ML random sequences w.r.t.  $P$ .

**Theorem 1** *If  $y^\infty \in \mathcal{R}^{P_Y}$  then  $P(\cdot|y^\infty)$  is a probability measure on  $\Omega$ .*

Remark: It is known that  $P(\cdot|y^\infty)$  is a probability measure on  $\Omega$  for almost all  $y^\infty$ .

Outline of proof)

- $\lim_{y \rightarrow y^\infty} P(x, y)/P_Y(y) \leq 1$  exists by martingale convergence theorem for random sequences.

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$$\begin{aligned} \forall x, 0 &= \lim_{y \rightarrow y^\infty} P(x|y) - P(x0|y) - P(x1|y) \\ &= P(x|y^\infty) - P(x0|y^\infty) - P(x1|y^\infty) \end{aligned}$$

qed

Lambalgen's theorem for correlated probability:

$$\mathcal{R}_{y^\infty}^P := \{x^\infty \in \Omega \mid (x^\infty, y^\infty) \in \mathcal{R}^P\}.$$

$P(\cdot \mid y^\infty)$  : conditional prob. w.r.t.  $y^\infty$ .

$\mathcal{R}^{P(\cdot \mid y^\infty), y^\infty} :=$  the set of ML random sequences w.r.t.  $P(\cdot \mid y^\infty)$ .

**Theorem 2** *Let  $P$  be a computable probability on  $X \times Y (= \Omega^2)$ . If  $P(\cdot \mid y^\infty)$  is computable relative to  $y^\infty \in \mathcal{R}^{PY}$ , then*

$$\mathcal{R}^{P(\cdot \mid y^\infty), y^\infty} = \mathcal{R}_{y^\infty}^P.$$

In [Takahashi 2008], “uniform computability” assumption was used to show  $\mathcal{R}^{P(\cdot \mid y^\infty), y^\infty} \supset \mathcal{R}_{y^\infty}^P$ .

Notation:  $S' \subset S$ ,  $\tilde{S}' := \cup_{s \in S'} \Delta(s)$ .

Outline of  $\mathcal{R}^{P(\cdot|y^\infty), y^\infty} \supset \mathcal{R}_{y^\infty}^P$  :

The proof is done by extending a ML-test w.r.t.  $P(\cdot|y^\infty)$  to a ML-test w.r.t.  $P$  on  $\Omega^2$ .

Fix  $y^\infty \in \mathcal{R}_Y^P$ .

$A, B$  : partial comp. func.

$A : S \times S \times \mathbb{N} \rightarrow \mathbb{Q}$ ,  $\forall x, k \exists y \sqsubset y^\infty, |P(x|y^\infty) - A(x, y, k)| < \frac{1}{k}$ .

$B : \mathbb{N} \times \mathbb{N} \times S \rightarrow S$ ,  $U_n^{y^\infty} := \{B(i, n, y) | \exists i, y \sqsubset y^\infty\}$  is a ML-test w.r.t.  $P(\cdot|y^\infty)$  and  $P(\tilde{U}_n^{y^\infty} | y^\infty) < 2^{-n}$ .

For simplicity, suppose that  $P(x|y^\infty) > 0$  for all  $x$ .

$$V_n := \{(x, y, k) \mid \exists i, B(i, n, y) = x, \frac{1}{k} < \frac{1}{2}A(x, y, k)\}.$$

Since  $P(x|y^\infty) > 0$ ,  $V_n$  is r.e. and  $P(x|y^\infty) < \frac{3}{2}A(x, y, k)$ .

Then there is a r.e.  $W_n \subset S \times S \times \mathbb{N}$  s.t.

$\tilde{W}'_n \subset \tilde{V}'_n$ , where  $W'_n = \{(x, y) \mid (x, y, k) \in W_n\}$ ,  $V'_n = \{(x, y) \mid (x, y, k) \in V_n\}$ ,

$$\sum_{(x,y,k) \in W_n} A(x, y, k)P(y) < 2^{-n+1},$$

$$\tilde{U}_n^{y^\infty} = \tilde{W}'_{n,y^\infty}.$$

Finally let

$$U_n := \{(x, z) \in S \times S \mid (x, y) \in W'_n, y \sqsubseteq z, \\ P(x, y) < \frac{3}{2}A(x, y, k)P(y)\}.$$

Then  $P(\tilde{U}_n) < \frac{3}{2} \sum_{(x,y) \in W'_n} A(x, y, k)P(y) < 2^{-n+2}$ .

Since  $P(x|y) \rightarrow P(x|y^\infty)$  as  $y \rightarrow y^\infty$  for  $y^\infty \in \mathcal{R}^{P_Y}$ ,

$$\tilde{U}_n^{y^\infty} = \tilde{U}_{n,y^\infty},$$

and  $U_n$  is a ML-test of  $P$ .

The converse inclusion is shown in [takahashi 2008]. ■

Monotone complexity

Order structure:

$$\Delta(x) := \{(\omega_1, \omega_2, \dots) \in \Omega^\infty \mid \forall i, x_i \sqsubseteq \omega_i\} \text{ for } x = (x_1, x_2, \dots) \in (S \cup \Omega)^\infty.$$

$$x \sqsubseteq y \text{ if } \Delta(x) \supset \Delta(y).$$

$((S \cup \Omega)^\infty, \sqsubseteq)$  : partially ordered set.

$$S^* := \cup_k S^k, \quad \eta : S^* \rightarrow S^\infty,$$

$$\eta(x) := (x, \lambda, \lambda, \dots)$$

$$[x] := \{y \mid \eta(x) = \eta(y), y \in S^*\}$$

$$[x] \sqsubseteq [y] \text{ if } \eta(x) \sqsubseteq \eta(y).$$

$S^*/\ker(\eta) := \{[x] \mid x \in S^*\}$  : partially ordered set

$\eta : S^*/\ker(\eta) \rightarrow (S \cup \Omega)^\infty, \quad \eta([x]) := \eta(x)$  : 1-1 order preserving.

$\bigvee A$ : the least upper bound of  $A$  for  $A \subset S^*/\ker(\eta)$ .  
The  $\bigvee A$  exists in  $(S \cup \Omega)^\infty$  iff  $\bigcap_{x \in A} \Delta(\eta(x)) \neq \emptyset$ .

Monotone function and complexity:

Let  $F \subset S^*/\ker(\eta) \times S^*/\ker(\eta)$  and  $F_p := \{x \mid (p, x) \in F\}$ .

Assume:

- a0)  $\forall p \in S^*/\ker(\eta), [\lambda] \in F_p$ .
- a1)  $\forall p \in S^*/\ker(\eta), \bigvee_{p' \sqsubseteq p} F_{p'}$  exists.

Set

$$f(p) := \bigvee_{p' \sqsubseteq p, p' \in S^*/\ker(\eta)} F_{p'} \text{ for } p \in (S \cup \Omega)^\infty. \quad (1)$$

Then

$$f : (S \cup \Omega)^\infty \rightarrow (S \cup \Omega)^\infty \text{ and } p' \sqsubseteq p \Rightarrow f(p') \sqsubseteq f(p).$$

Conversely, let  $f : (S \cup \Omega)^\infty \rightarrow (S \cup \Omega)^\infty$  be a monotone function, and set

$$F := \{(p, x) \in S^* \times S^* \mid \eta(x) \sqsubseteq f(\eta(p))\},$$

Then  $\forall F_p = f(p)$ ,

$F$  satisfies a0 and a1, and the function defined by  $F$  coincides with  $f$ .

If  $F$  is a r.e. set and  $F$  satisfies a0 and a1, then the function  $f$  defined by (1) is called *computable monotone function*.

$$Km_f(x) := \min\{|p| \mid x \sqsubseteq f(p)\}, \text{ for } x \in (S \cup \Omega)^\infty, |p| := \sum_i |p_i|.$$

$$Km_f(x) := Km_f(\eta(x)), \text{ for } x \in S^*.$$

$$Km(x) := Km_u(x), \text{ } u \text{ is optimal.}$$

Levin-Schnorr theorem for product space:

$$\mathcal{A} \subset S^* \text{ and r.e..}$$

Condition 1) if  $x, y \in \mathcal{A}$  then,  $x$  and  $y$  are comparable or

$$\Delta(x) \cap \Delta(y) = \emptyset.$$

Condition 2) there is a recursive function  $f : S^* \times \mathbb{N} \rightarrow \mathcal{A}$  such that for any  $x \in S^*$ ,  $\Delta(x) = \widetilde{f(x, \mathbb{N})}$  and  $f(x, \mathbb{N})$  is prefix-free.

$$\mathcal{A}(x^\infty) := \{x \in S^* \mid x \in \mathcal{A}, x \sqsubset x^\infty\}.$$

$P$  : computable probability on  $(\mathcal{B}_{\Omega^\infty}, \Omega^\infty)$ .

$$\mathcal{B}_{\Omega^\infty} = \sigma\{\Delta(\eta(x)) \mid x \in S^*\}.$$

**Theorem 3 (Levin-Schnorr theorem on product space)** *Under Condition 1 and 2,*

$$x^\infty \in \mathcal{R}^P \text{ iff } \sup_{x \in \mathcal{A}(x^\infty)} -\log P(x) - Km(x) < \infty.$$

If  $u : (S \cup \Omega)^\infty \rightarrow (S \cup \Omega)^\infty$  optimal and  $f : S \cup \Omega \rightarrow (S \cup \Omega)^\infty$  then

$$Km_u(x) \leq Km_f(x) + O(1).$$

### **Corollary 1 (One dimensional coding)**

*If  $\mathcal{A} \subset S^*$  is decidable*

*then there is  $f : S \cup \Omega \rightarrow (S \cup \Omega)^\infty$  and*

$$Km_f(x) \leq -\log P(x) + O(1) \text{ for } x \in \mathcal{A}.$$

*In particular if  $x^\infty \in \mathcal{R}^P$  and  $\mathcal{A}$  is decidable then under Condition 1 and 2,*

$$\sup_{x \in \mathcal{A}(x^\infty)} |Km_u(x) - Km_f(x)| < \infty.$$

Consistency of posterior distribution:

Example: Let  $P_\theta$  be Bernoulli model, then  $\lim_n E(\theta|X_1, X_2, \dots, X_n) = \theta$ ,  $P_\theta - a.s.$  (consistency of posterior distribution).

We have a stronger statement for the consistency of posterior distribution.

**Theorem 4** *Let  $P$  be a computable probability on  $X \times Y$ , where  $X = Y = \Omega$ . Assume that  $P_Y(y) > 0$  for all  $y \in S$ . The following six statements are equivalent:*

(1)  $P(\cdot | y) \perp P(\cdot | z)$  if  $\Delta(y) \cap \Delta(z) = \emptyset$  for  $y, z \in S$ .

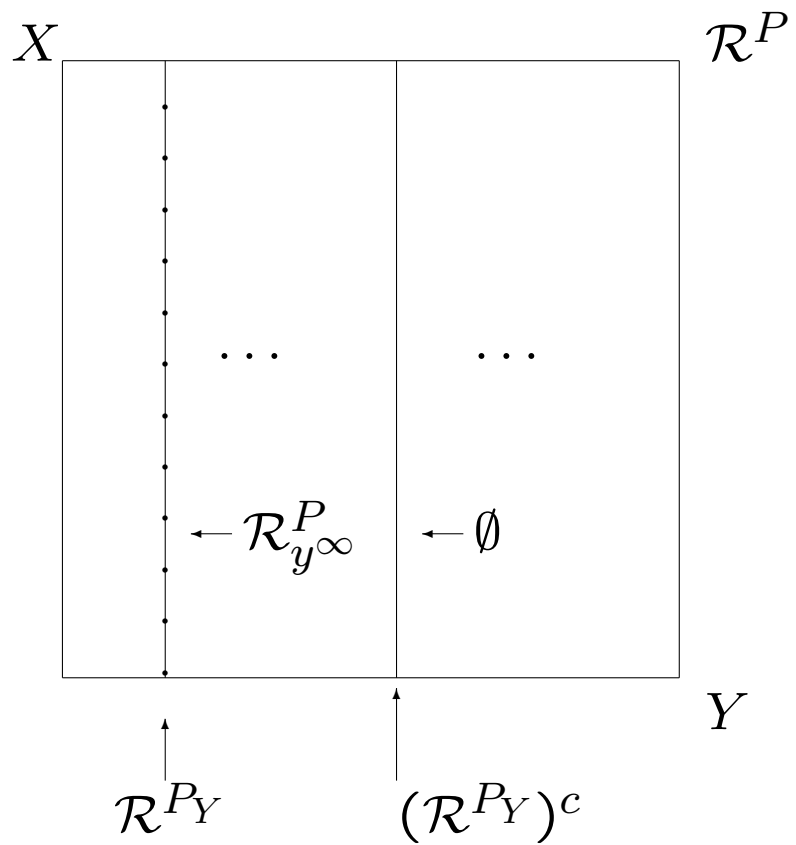
(2)  $\mathcal{R}^{P(\cdot | y)} \cap \mathcal{R}^{P(\cdot | z)} = \emptyset$  if  $\Delta(y) \cap \Delta(z) = \emptyset$  for  $y, z \in S$ .

(3)  $P_{Y|X}(\cdot | x)$  converges weakly to  $I_{y^\infty}$  as  $x \rightarrow x^\infty$  for  $(x^\infty, y^\infty) \in \mathcal{R}^P$ , where  $I_{y^\infty}$  is the probability that has probability of 1 at  $y^\infty$ .

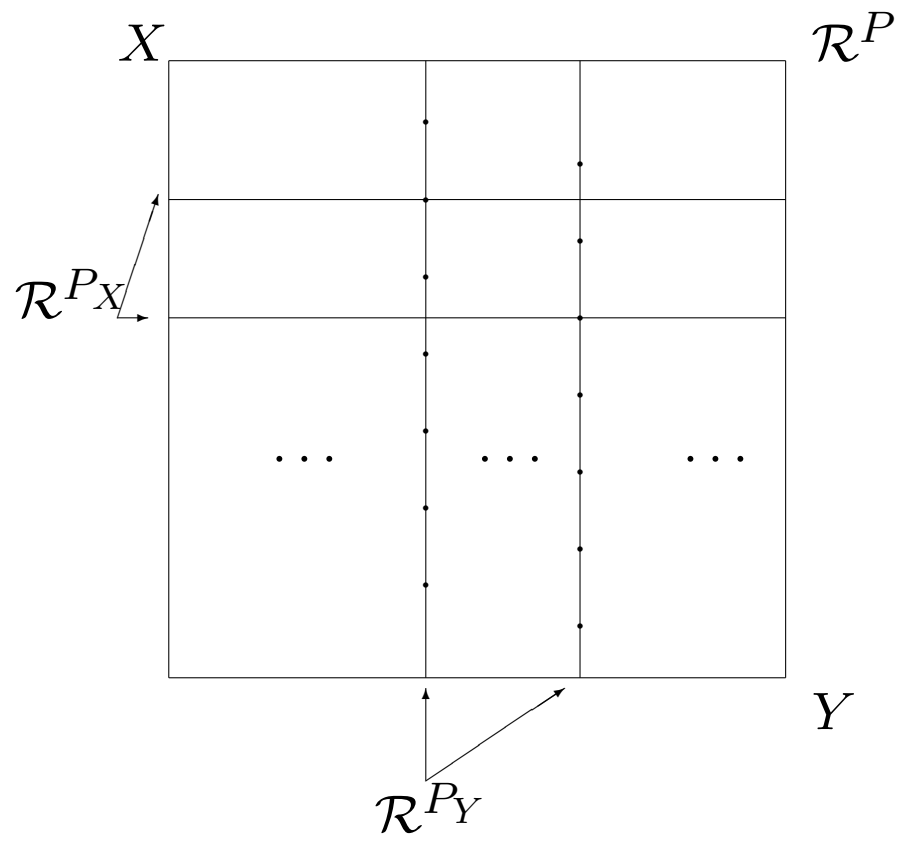
(4)  $\mathcal{R}_{y^\infty}^P \cap \mathcal{R}_{z^\infty}^P = \emptyset$  if  $y^\infty \neq z^\infty$ .

(5) There exists  $f : X \rightarrow Y$  such that  $f(x^\infty) = y^\infty$  for  $(x^\infty, y^\infty) \in \mathcal{R}^P$ .

(6) There exists  $f : X \rightarrow Y$  and  $Y' \subset Y$  such that  $P_Y(Y') = 1$  and  $f = y^\infty, P(\cdot; y^\infty) - a.s.$  for  $y^\infty \in Y'$ .



General



Consistent

## References:

Hayato Takahashi, On a definition of random sequences with respect to conditional probability, Information and Computation, 2008.

Thank you.