

# Muchnik degrees and Medvedev degrees of the randomness notions

Kenshi Miyabe (宮部賢志) @ Meiji University

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## Background

- ❖ Randomness hierarchy
- ❖ Mass problems
- ❖ Main results

Proof

Further results

# Background

# Randomness hierarchy

$$\text{KR} \supset \text{SR} \supset \text{CR} \supset \text{MLR} \supset \text{DiffR} \begin{array}{l} \supset \text{W2R} \\ \supset \text{DemR} \end{array} \supset \text{2R}$$

We know  $\exists X \in \text{SR} \setminus \text{CR}$ .

Given a Schnorr random, can we compute a computably random?

# Mass problems

**Definition 1.** Let  $P, Q \subseteq 2^\omega$ .

$P$  is **Muchnik reducible** to  $Q$  ( $P \leq_w Q$ ) if, for every  $f \in Q$ , there exists  $g \in P$  such that  $g \leq_T f$ .

$P$  is **Medvedev reducible** to  $Q$  ( $P \leq_s Q$ ) if, there exists a Turing functional  $\Phi$  such that  $\Phi^f \in P$  for every  $f \in Q$ .

The difference is uniformity.

# Main results

## Muchnik degrees

$$\text{KR} <_w \text{SR} \equiv_w \text{CR} <_w \text{MLR} \equiv_w \text{DiffR} <_w \begin{matrix} \text{W2R} \\ \text{DemR} \end{matrix} <_w \begin{matrix} \text{2R} \\ \end{matrix}$$

## Medvedev degrees

$$\text{SR} <_s \text{CR}, \text{MLR} <_s \text{DiffR}$$

## Background

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## Proof

$$\diamond \text{SR} \equiv_w \text{CR}$$

$$\diamond \text{SR} <_s \text{CR}$$

$$\diamond \text{SR} \supsetneq \text{CR}$$

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$$\diamond \text{SR} <_s \text{CR}$$

$$\diamond \text{CR}(\mu) \not\subseteq \text{CR}(\lambda)$$

$$\diamond \text{CR}(\mu) \subseteq \text{CR}(\lambda)$$

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## Further results

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# Proof

$$\text{SR} \equiv_w \text{CR}$$

Let  $X \in \text{SR}$ .

If  $X$  is not high,  $X$  is already in CR.

If  $X$  is high,  $X$  can compute a real in CR.

Both from Nies, Stephan, and Terwijn (2005).

$$SR <_s CR$$

The goal is

$$\forall \Phi \exists X \in SR [\Phi(X) \notin CR].$$

When  $\Phi = \text{id}$ , this means

$$X \in SR \setminus CR.$$

Thus, we extend the method separating SR and CR.



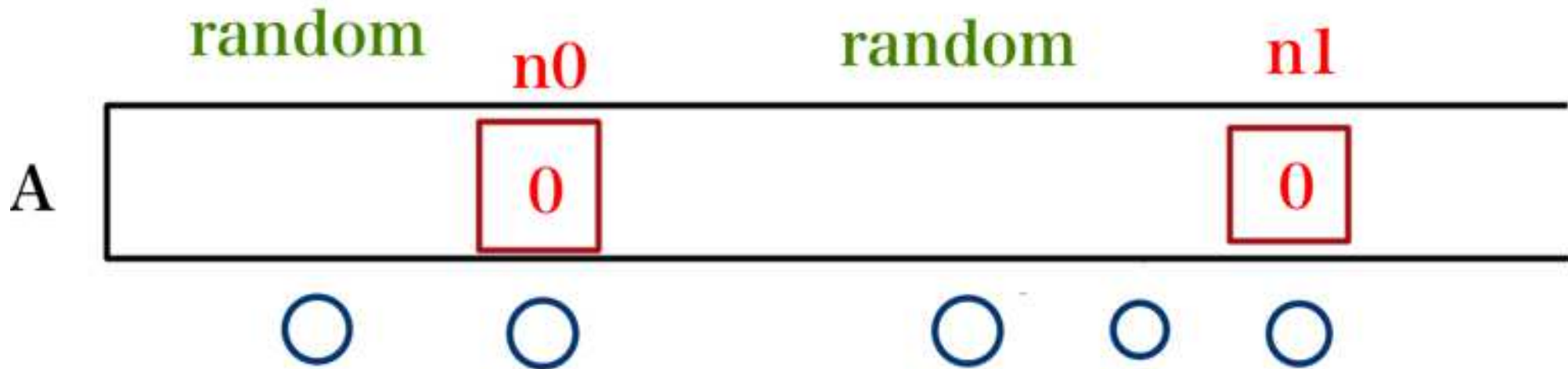
# SR $\not\subseteq$ CR

id case

- Construct a random set  $A$ .
- Forcing  $A(n_k) = 0$  in sparse positions  
 $\Rightarrow$  too sparse not to be Schnorr random
- The number of candidates of  $n_k$  is small  
 $\Rightarrow$  so small that some computable martingale succeeds on it.

$$SR \not\subseteq CR$$

The positions forcing 0 are sparse.



The numbers of candidates are small.

# SR $<_s$ CR

general case

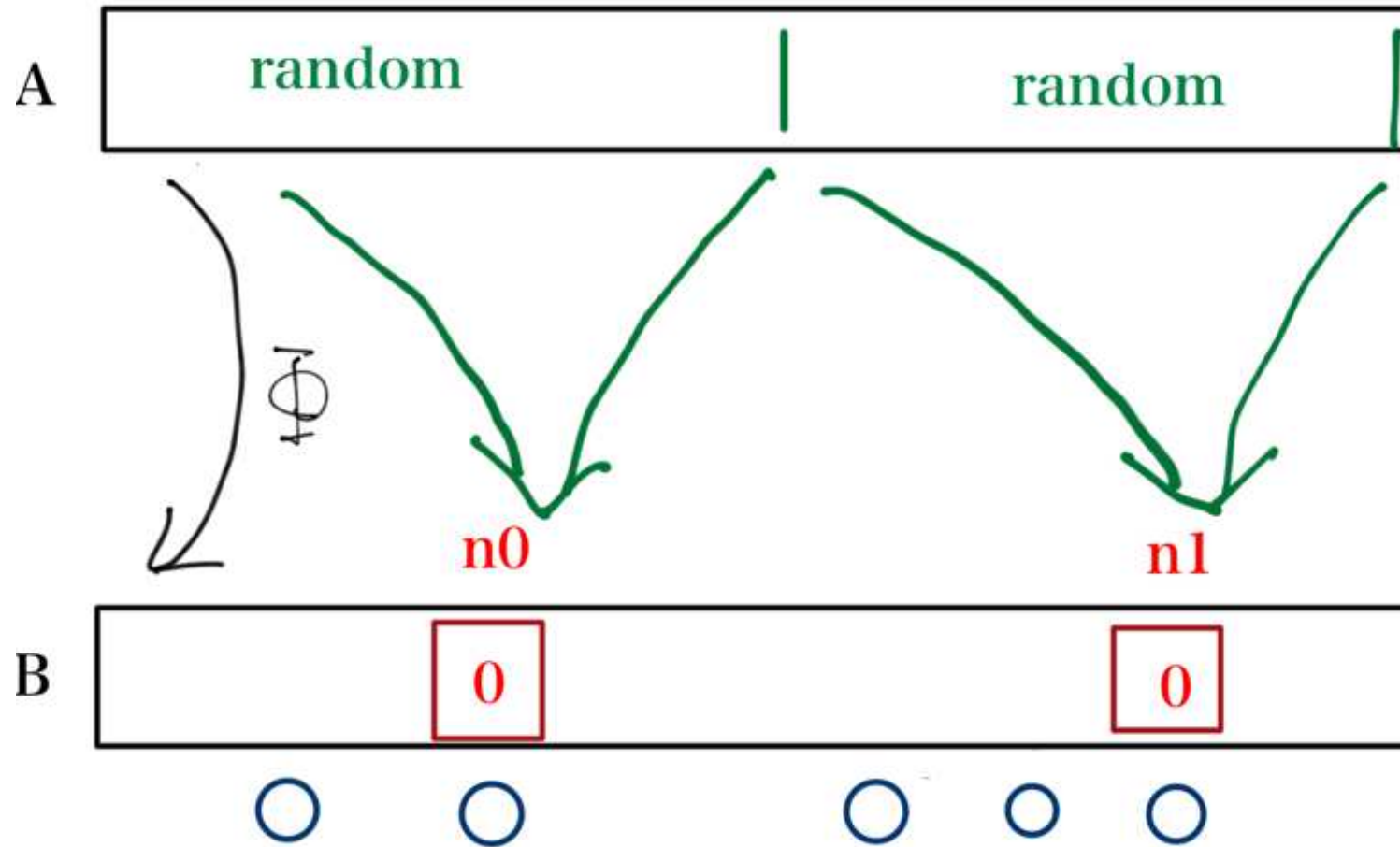
- Construct  $A \in \text{SR}$  and  $B = \Phi(A) \notin \text{CR}$ .
- Forcing  $B(n_k) = 0$  in some positions (\*)
- The number of candidates of  $n_k$  should be small  
 $\Rightarrow B \notin \text{CR}$ .

The requirement (\*) may be strong because

$$\lambda(\{X \in 2^\omega : \Phi(X)(n_k) = 0\})$$

may be too small (even empty).

$$SR <_s CR$$



The numbers of candidates are small.

# SR $<_s$ CR

We divide the case into two by the induced measure  $\mu$ .

- $\mu$  is "close to" uniform measure ( $\text{CR}(\mu) \subseteq \text{CR}(\lambda)$ )  
 $\Rightarrow$  the same method can be applied
- $\mu$  is "far from" uniform measure ( $\text{CR}(\mu) \not\subseteq \text{CR}(\lambda)$ )  
 $\Rightarrow$  we can show it by a different reason

$$\text{CR}(\mu) \not\subseteq \text{CR}(\lambda)$$

$$\exists Y \in \text{CR}(\mu) \setminus \text{CR}(\lambda)$$

By no-randomness-from-nothing for CR,

$$\exists X \in \text{CR} [\Phi(X) = Y].$$

Then,  $X \in \text{SR}$  and  $\Phi(X) \notin \text{CR}$ .

$$\text{CR}(\mu) \subseteq \text{CR}(\lambda)$$

**Lemma 2** (essentially Bienvenu-Merkle). *Let  $\mu, \nu$  be computable measures.*

$$\text{CR}(\mu) \subseteq \text{CR}(\nu) \Rightarrow \text{MLR}(\mu) \subseteq \text{MLR}(\nu) \Rightarrow \mu \ll \nu$$

$\ll$  *means absolute continuity.*

$$\text{CR}(\mu) \subseteq \text{CR}(\lambda)$$

**Lemma 3.** *Let  $\Phi$  be an a.e. computable function. Let  $\mu$  be the induced measure from  $\Phi$  and  $\lambda$ . Assume  $\lambda \ll \mu$ . For each  $\sigma \in 2^{<\omega}$ ,*

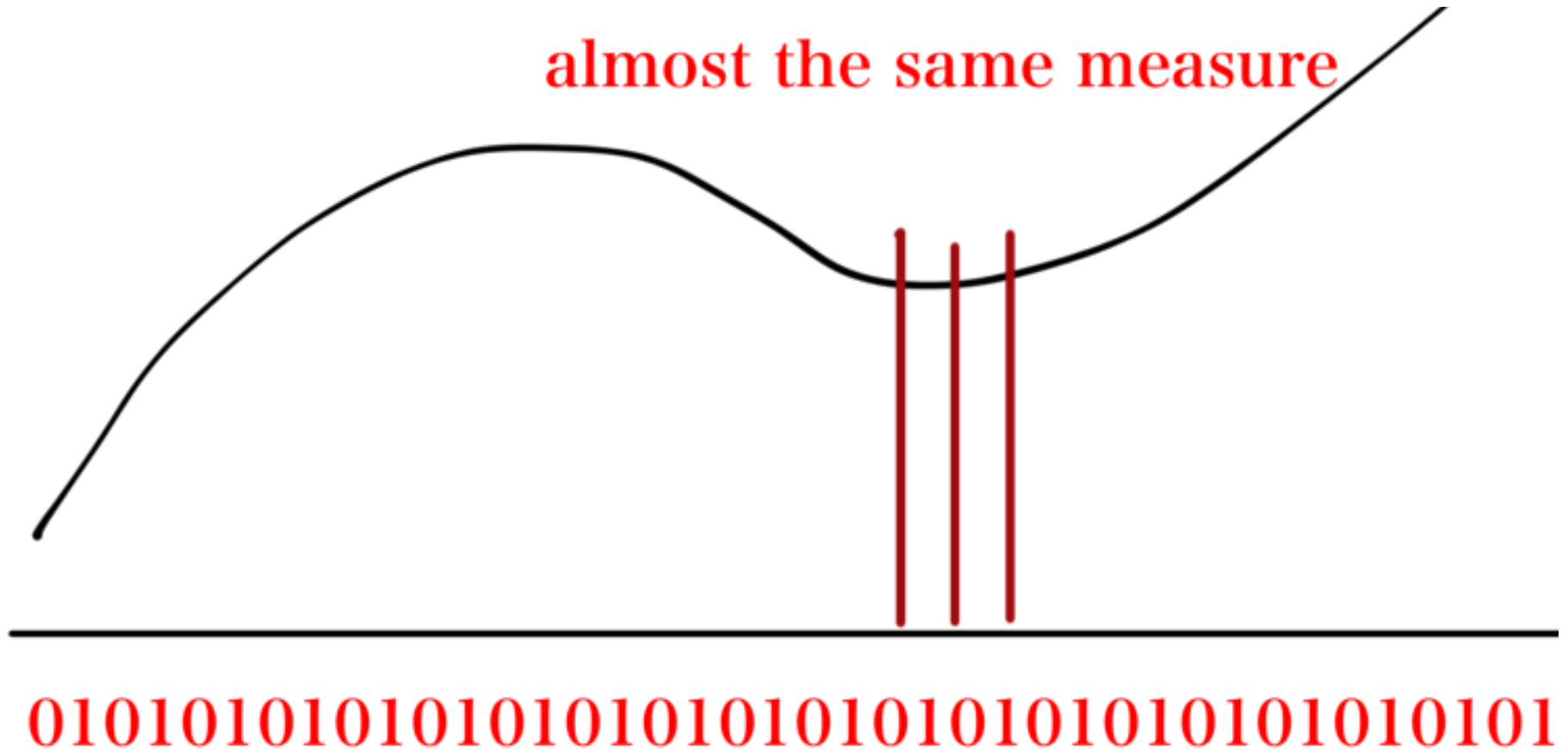
$$\lim_{n \rightarrow \infty} \lambda\{X \in [\sigma] : \Phi(X)(n) = 0\} = \frac{1}{2}\lambda(\sigma).$$

*Proof.* By the Radon-Nikodym theorem and Lévy's zero-one law. □



$$\text{CR}(\mu) \subseteq \text{CR}(\lambda)$$

almost the same measure



Background

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Proof

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**Further results**

- ❖ Question
- ❖ m-degree
- ❖ Summary

# Further results

# Question

**Question 4.** Does there exist  $A \in \text{SR}$  such that, if  $B \leq_{tt} A$  then  $B \notin \text{CR}$ .

How about wtt?

I conjecture that we can not tt-compute (or wtt-compute) a computably random from a Schnorr random even nonuniformly.

# ***m-degree***

**Definition 5.**  $X \leq_m Y$  if  $\exists$  comp.  $f$  such that

$$n \in X \iff f(n) \in Y$$

**Theorem 6.** *There exists  $A \in \text{SR}$  such that, if  $B \leq_m A$  then  $B \notin \text{CR}$ .*

Every computable subsequence of  $A \in \text{SR}$  is not computably random.

So, regularity prevails anywhere!

The proof is a slight extension of the id case.

# *Summary*

- We found two problems that is possible non-uniformly but not possible uniformly.
- Analytical tools are useful to show results in computability. In particular, a.e. computable functions can be studied more from the measure-theoretic perspective.