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# AN OPEN PROBLEM REGARDING THE CONVERGENCE OF UNIVERSAL A PRIORI PROBABILITY

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## Abstract

Is the textbook result that Solomonoff's universal posterior converges to the true posterior for all Martin-Löf random sequences true?

**Universal induction.** Induction problems can be phrased as sequence prediction tasks. This is, for instance, obvious for time series prediction, but also includes classification tasks. Having observed data  $x_t$  at times  $t < n$ , the task is to predict the  $t$ -th symbol  $x_t$  from sequence  $x = x_1 \dots x_{t-1}$ . The key concept to attack general induction problems is Occam's razor and to a less extent Epicurus' principle of multiple explanations. The former/latter may be interpreted as to keep the simplest/all theories consistent with the observations  $x_1 \dots x_{t-1}$  and to use these theories to predict  $x_t$ . Solomonoff [Sol64, Sol78] formalized and combined both principles in his universal prior  $M(x)$  which assigns high/low probability to simple/complex environments, hence implementing Occam and Epicurus.  $M(x)$  is defined as the probability that a universal Turing machine  $U$  outputs a string starting with  $x$ , when provided with fair coin flips on the input tape.

**Posterior convergence.** Solomonoff's [Sol78] central result is that if the probability  $\mu(x_t|x_1 \dots x_{t-1})$  of observing  $x_t$  at time  $t$ , given past observations  $x_1 \dots x_{t-1}$  is a computable function, then the universal posterior  $M_t := M(x_t|x_1 \dots x_{t-1})$  converges (rapidly!) *with  $\mu$ -probability 1* (w.p.1) for  $t \rightarrow \infty$  to the true posterior  $\mu_t := \mu(x_t|x_1 \dots x_{t-1})$ , hence  $M$  represents a universal predictor in case of unknown  $\mu$ . Convergence of  $M_t$  to  $\mu_t$  w.p.1 tells us that  $M_t$  is close to  $\mu_t$  for sufficiently large  $t$  for "almost all" sequences  $x_{1:\infty}$  (we abbreviate  $x_{1:n} := x_1 \dots x_n$ ). It says nothing about whether convergence is true for any *particular* sequence (of measure 0).

**Martin-Löf randomness** is the standard notion to capture convergence for individual sequences and is closely related to Solomonoff's universal prior. Levin gave a characterization equivalent to Martin-Löf's (M.L.) original definition [Lev73]:

*A sequence  $x_{1:\infty}$  is  $\mu$ -random (in the sense of M.L.) iff there is a constant  $c$  such that  $M(x_{1:n}) \leq c \cdot \mu(x_{1:n})$  for all  $n$ .*

One can show that a  $\mu$ -random sequence  $x_{1:\infty}$  passes *all* thinkable effective randomness tests, e.g. the law of large numbers, the law of the iterated logarithm, etc. In particular, the set of all  $\mu$ -random sequences has  $\mu$ -measure 1.

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\*A prize of 128 Euro for a solution of this problem is offered.

**Open problem.** An interesting open question is whether  $M_t$  converges to  $\mu_t$  (in difference or ratio) individually for all Martin-Löf random sequences. Clearly, Solomonoff’s result shows that convergence may at most fail for a set of sequences with  $\mu$ -measure zero. A convergence result for  $\mu$ -random sequences is particularly interesting and natural in this context, since  $\mu$ -randomness can be defined in terms of  $M$  itself (see above).

**Proof attempts.** Attempts to convert the convergence results w.p.1 to effective  $\mu$ -randomness tests fail, since  $M_t$  is not lower semi-computable. In [LV97, Th.5.2.2] and [VL00, Th.10] the following Theorem is stated:

“Let  $\mu$  be a positive recursive measure. If the length of  $y$  is fixed and the length of  $x$  grows to infinity, then  $M(y|x)/\mu(y|x) \rightarrow 1$  with  $\mu$ -probability one. The infinite sequences  $\omega$  with prefixes  $x$  satisfying the displayed asymptotics are precisely [ $\Rightarrow$ ] and [ $\Leftarrow$ ] the  $\mu$ -random sequences.”

While convergence w.p.1 is correct if appropriately interpreted,<sup>1</sup> the proof that convergence holds for  $\mu$ -random sequences is incomplete: “ $M(x_{1:n}) \leq c \cdot \mu(x_{1:n}) \forall n \Rightarrow \lim_{n \rightarrow \infty} M(x_{1:n})/\mu(x_{1:n})$  exists” has been used, but not proven, and may indeed be wrong. Vovk [Vov87] shows that for two finitely computable semi-measures  $\mu$  and  $\rho$ , and  $x_{1:\infty}$  being  $\mu$ - and  $\rho$ -random that  $\rho_t/\mu_t \rightarrow 1$ . If  $M$  were recursive, then this would imply  $M_t/\mu_t \rightarrow 1$  for every  $\mu$ -random sequence  $x_{1:\infty}$ , since every sequence is  $M$ -random. Since  $M$  is not recursive Vovk’s theorem cannot be applied and it is not obvious how to generalize it. So the question of individual convergence remains open.

**Conclusions.** Contrary to what was believed before, the question of posterior convergence  $M_t/\mu_t \rightarrow 1$  (also  $M_t \rightarrow \mu_t$ ) for all  $\mu$ -random sequences is still open. In [Hut03] we introduce a new flexible notion of randomness which contains Martin-Löf randomness as a special case. This notion is used to show that standard proof attempts of  $M_t/\mu_t \xrightarrow{M.L.} 1$  based on so called dominance only must fail, indicating that this problem may be a hard one.

## References.

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<sup>1</sup>The formulation of the Theorem is quite misleading in general: First, for off-sequence  $y$  convergence w.p.1 does not hold ( $xy$  must be demanded to be a prefix of  $x_{1:\infty}$ ). Second, the proof of [ $\Leftarrow$ ] has loopholes (see main text). Last, [ $\Rightarrow$ ] is given without proof and is probably wrong. Also the assertion in [LV97, Th.5.2.1] that  $S_t$  converges to zero faster than  $1/t$  cannot be made, since  $S_t$  may not decrease monotonically.