

Universality, optimality, and randomness deficiency

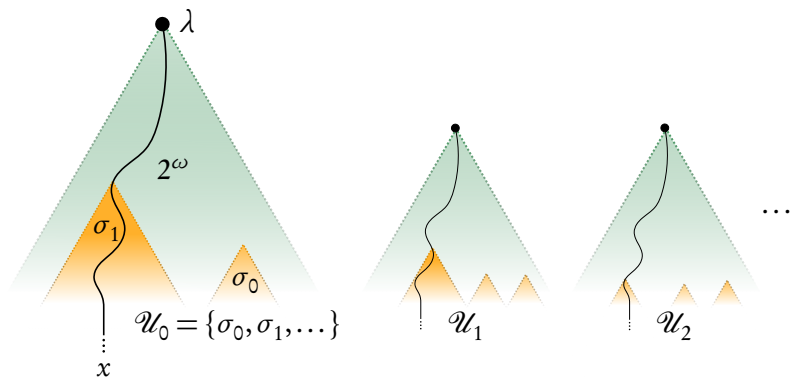
Rupert Hölzl



Universität der Bundeswehr München

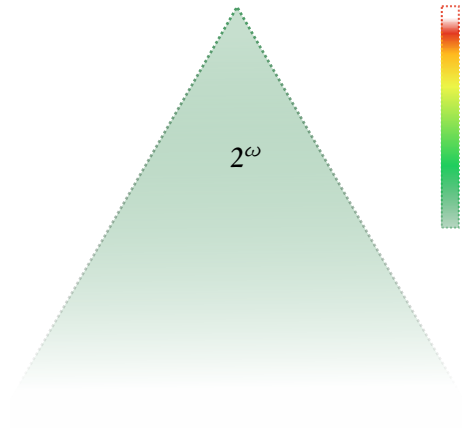
Joint work with Paul Shafer

Reminder: Martin-Löf randomness

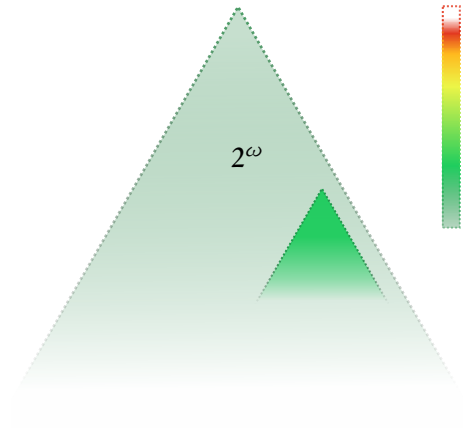


- 1** **Martin-Löf randomness.** Real is not random if in the intersection of an effective sequence of c.e. classes, whose measure tends to 0 at a guaranteed minimum speed.

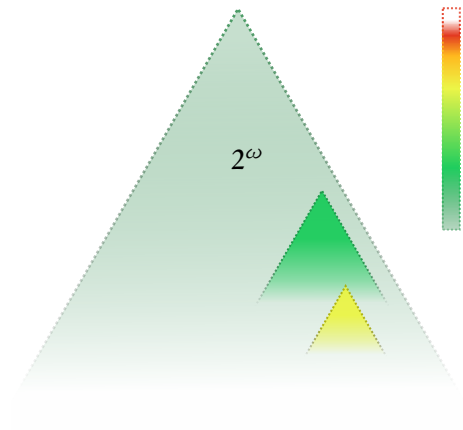
Non-randomness heatmap



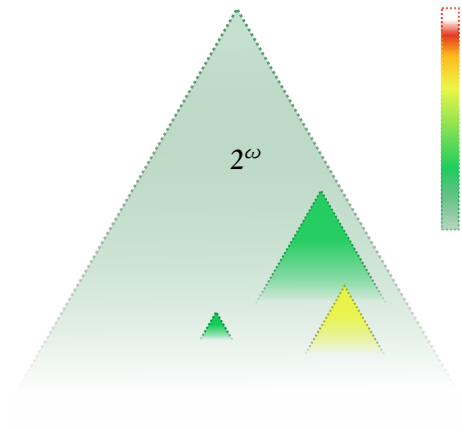
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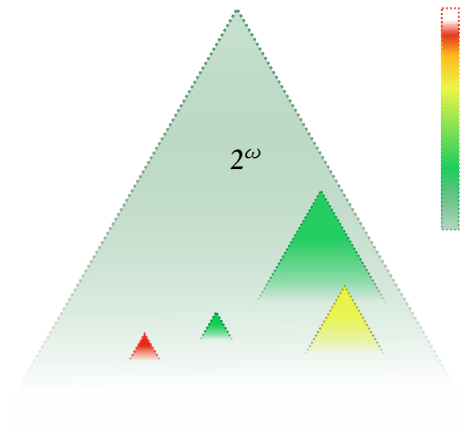
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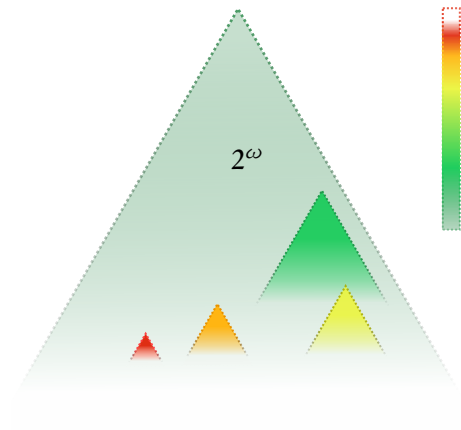
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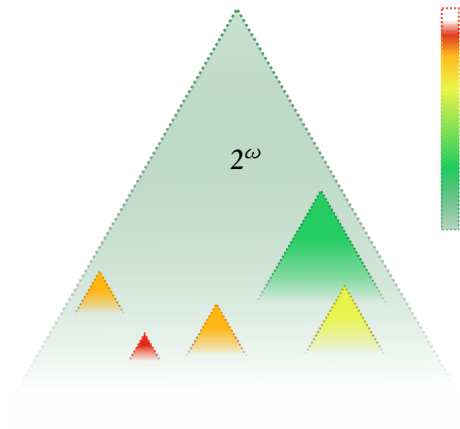
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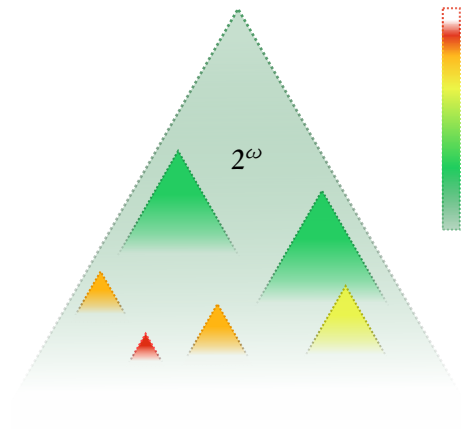
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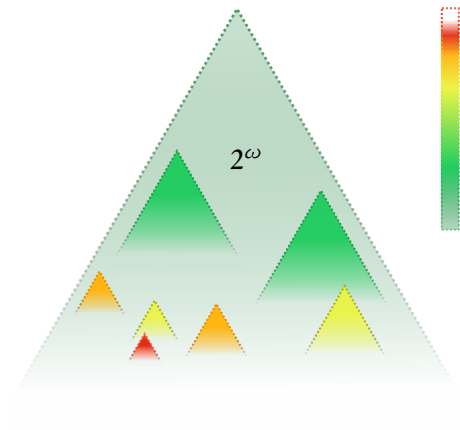
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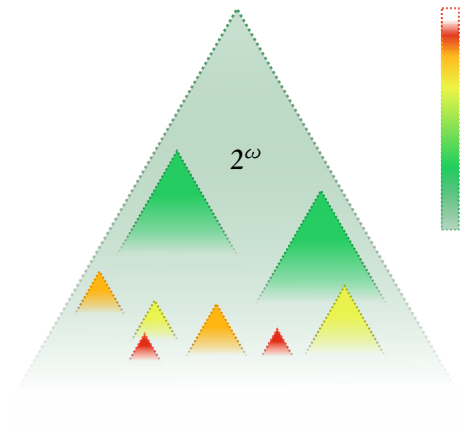
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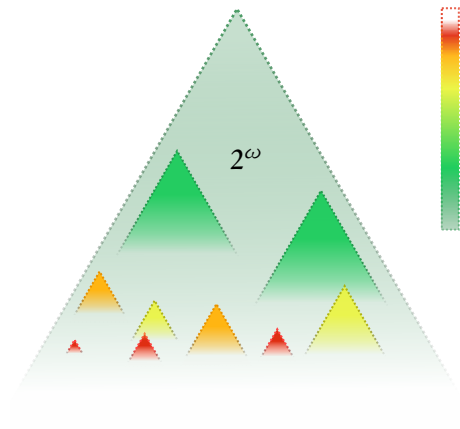
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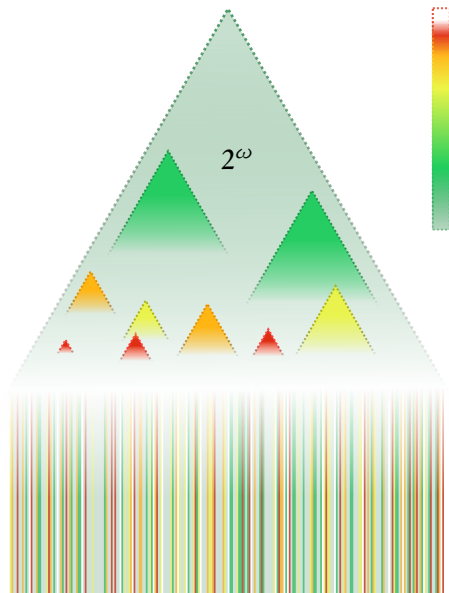
Non-randomness heatmap



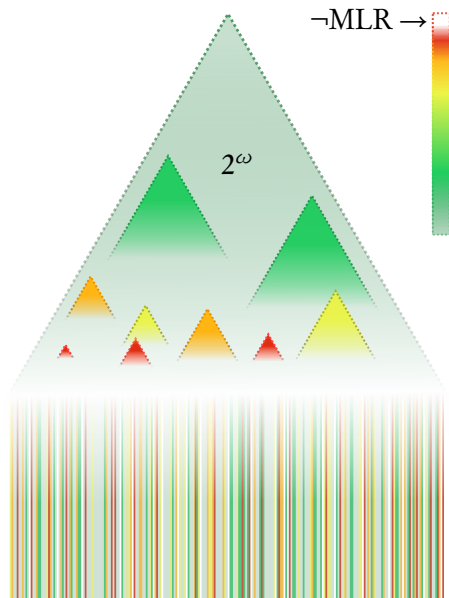
Non-randomness heatmap



Non-randomness heatmap



Non-randomness heatmap



Main question

- 1 The picture gives a visual idea of the interna of Martin-Löf tests.
(In general, of course, only universal tests are interesting.)
- 2 But there are many different universal ML-tests.
Are their structures always similar?
Or can they be very different?
If yes, how different?
- 3 More precisely:
How *computationally* different can they be?

Three approaches

- 1 We will study this question using tools of increasing precision.
 - Classical computability
 - Layerwise computability
 - Weihrauch degrees

1

Preliminaries

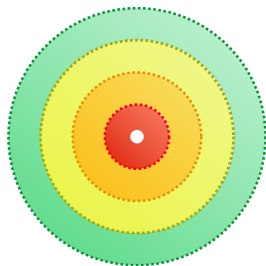
Types of Martin-Löf tests

- 1 Definition.** A Martin-Löf test \mathcal{U} is *universal* if, for every Martin-Löf test \mathcal{V} ,

$$\bigcap_{i \in \omega} \mathcal{V}_i \subseteq \bigcap_{i \in \omega} \mathcal{U}_i.$$

- 2 Definition.** A Martin-Löf test \mathcal{U} is *optimal* if, for every Martin-Löf test \mathcal{V} ,

$$\exists c \forall i: \mathcal{V}_{i+c} \subseteq \mathcal{U}_i.$$



- 1 **Definition.** A Martin-Löf test \mathcal{V} is *nested* if, for all i ,

$$\mathcal{V}_{i+1} \subseteq \mathcal{V}_i.$$

- 2 In this talk, we implicitly assume that all tests are nested, in particular when presenting intuitive arguments and proof sketches. All results also hold without that assumption.

- 1 Definition (Martin-Löf).** If $X \in \text{MLR}$ and \mathcal{U} is a universal ML-test, then the *randomness deficiency of X relative to \mathcal{U}* is defined as

$$\text{rd}_{\mathcal{U}}(X) = \min\{i : X \notin \mathcal{U}_i\}.$$

- 2 Intuition.** “The color of X .”
(Strictly speaking “the color of X plus 1.”)

2

Classical computability

Strongly non-optimal universal tests

- 1 **Fact.** If \mathcal{U} is an ML-test such that, for every ML-test \mathcal{V} ,

$$\forall i \exists j: \mathcal{V}_j \subseteq \mathcal{U}_i,$$

then \mathcal{U} is universal.

- 2 **Theorem.** The inverse is false. That is, there exist ML-tests \mathcal{U} and \mathcal{V} such that \mathcal{U} is universal but

$$\exists i \forall j: \mathcal{V}_j \not\subseteq \mathcal{U}_i.$$

(\mathcal{V} can even be chosen optimal.)

- 3 **In other words.**

- A function $f: \omega \rightarrow \omega$ such that $\forall i: \mathcal{V}_{f(i)} \subseteq \mathcal{U}_i$ need not exist.
- That is, there are strongly non-optimal universal tests.

Non-optimal tests and incomputability

1 Theorem. There are ML-tests \mathcal{W} and \mathcal{V} such that

- \mathcal{W} is universal,
- $\forall i \exists j: \mathcal{V}_j \subseteq \mathcal{W}_i$,
- if $f: \omega \rightarrow \omega$ satisfies $\forall i: \mathcal{V}_{f(i)} \subseteq \mathcal{W}_i$, then $f \geq_T 0''$.

(\mathcal{V} can even be chosen optimal.)

2 Summary.

- There are strongly non-optimal universal tests.
- Even when functions $f: \omega \rightarrow \omega$ such that $\forall i: \mathcal{V}_{f(i)} \subseteq \mathcal{U}_i$ exist, they may be strongly incomputable.

3

Layerwise computability

Layerwise computability

- 1 Let \mathcal{U} be a universal ML-test.
- 2 **Definition.** A function f is *exactly \mathcal{U} -layerwise computable* if there is a Turing functional Φ such that

$$\forall X \in \text{MLR}: \Phi(\langle X, \text{rd}_{\mathcal{U}}(X) \rangle) = f(X).$$

- 3 **Intuition.** “If you know the color of X you can compute $f(X)$.”
- 4 **Definition (Hoyrup and Rojas).** A function f is *\mathcal{U} -layerwise computable* if there is a Turing functional Φ such that

$$\forall i \forall X \in 2^\omega \setminus \mathcal{U}_i: \Phi(\langle X, i \rangle) = f(X).$$

- 5 **Different tests give different colors to X .**
How does this influence the computability of $f(X)$?

Layerwise computability depends on the test

- 1 Theorem (Hoyrup and Rojas).** Let \mathcal{U} be an *optimal* ML-test, and let $\mathcal{A} \subseteq 2^\omega$. Then \mathcal{A} is effectively measurable if and only if $\mathbb{1}_{\mathcal{A}}$ is \mathcal{U} -layerwise computable.
- 2 Theorem.** There is an effectively measurable set $\mathcal{A} \subseteq 2^\omega$ and a universal ML-test \mathcal{W} such that $\mathbb{1}_{\mathcal{A}}$ is not \mathcal{W} -layerwise computable.
- 3 Corollary.** Layerwise computability depends on the test.
(That is, there are universal ML-tests \mathcal{U} and \mathcal{W} and a function that is \mathcal{U} -layerwise computable but not \mathcal{W} -layerwise computable.)

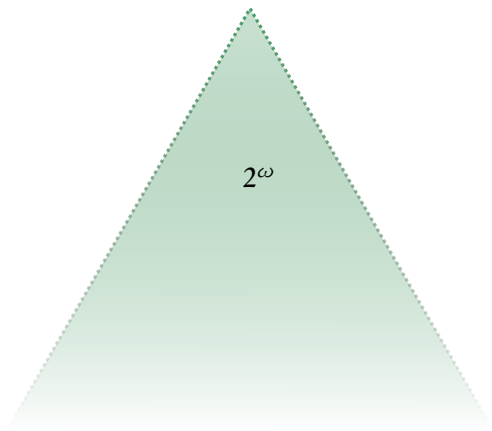
Layerwise computable vs. *exactly* layerwise computable

- 1 Theorem.** Let \mathcal{U} be a universal ML-test. Then $\text{rd}_{\mathcal{U}}$ is not \mathcal{U} -layerwise computable.
- 2 Corollary.** For every universal ML-test \mathcal{U} there is a function that is exactly \mathcal{U} -layerwise computable but not \mathcal{U} -layerwise computable.
- 3 Summary.** The class of exactly layerwise computable functions is *always* strictly larger than the class of layerwise computable functions.

Exact layerwise computability depends on the test

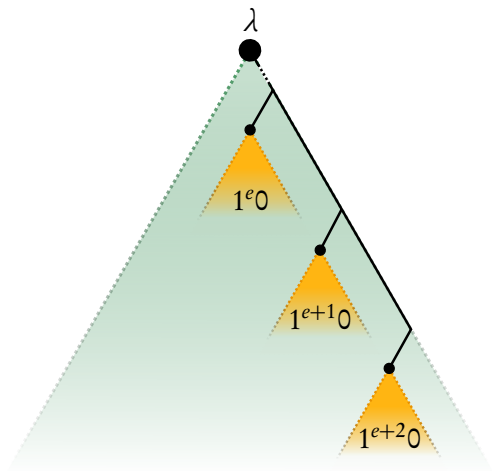
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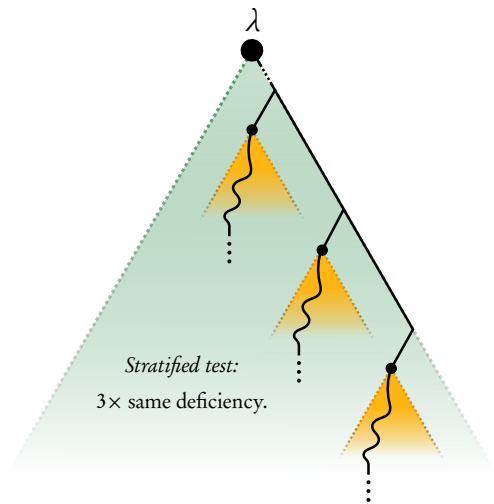
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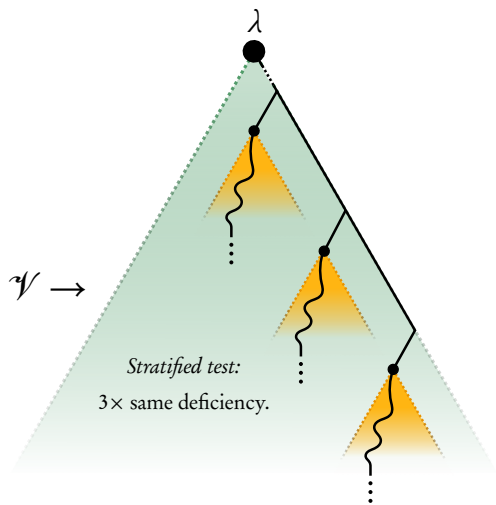
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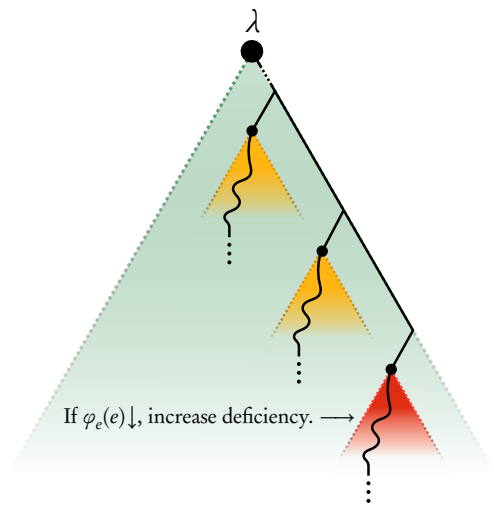
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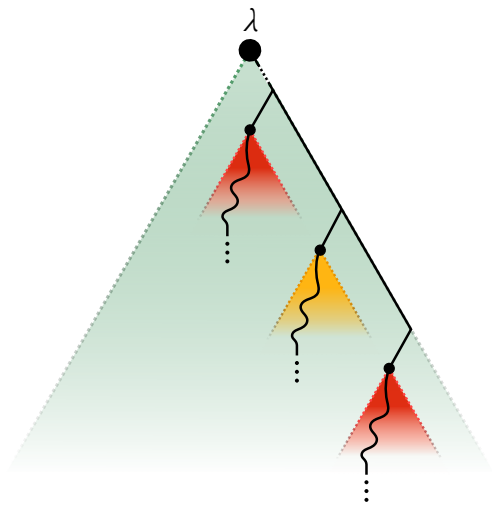
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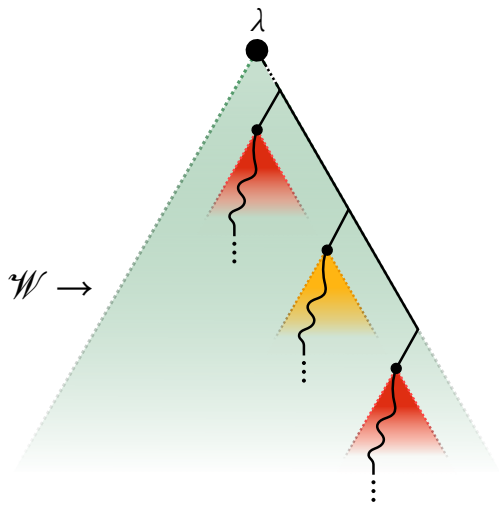
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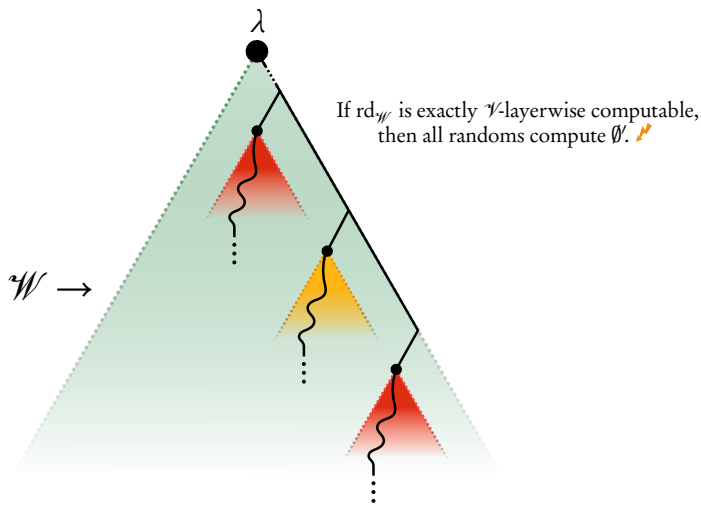
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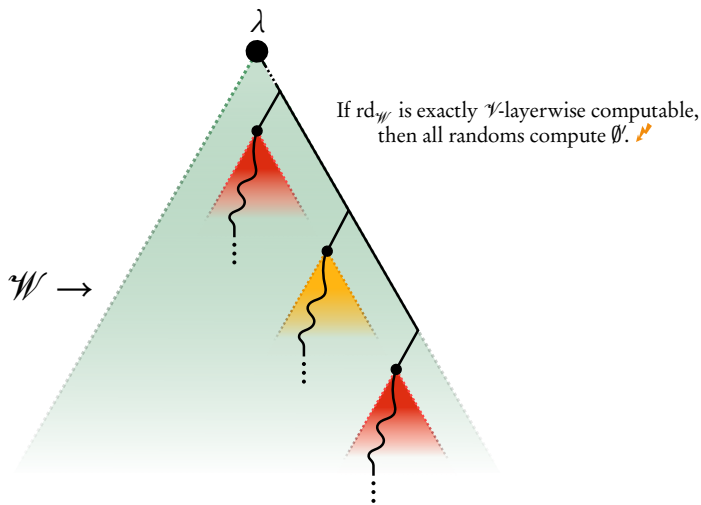
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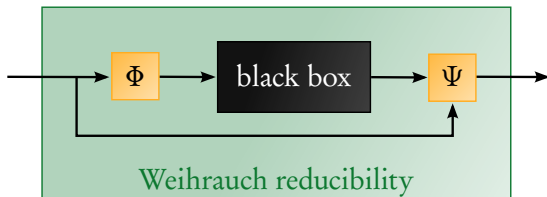
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Weihrauch degrees

Reminder: Weihrauch degrees

- 1 **General idea.** We look at mathematical tasks where a problem is given to a black box, and the black box has to give back a solution to the problem.
- 2 **Example.** Given a bounded sequence of rationals, the black box has to produce an accumulation point.
- 3 **Reducibility.** Assume we have a black box solving a certain problem \mathcal{A} . Can we use it to solve another problem \mathcal{B} ?
- 4 **Approach.**
 - Code an instance B of problem \mathcal{B} into a valid instance A of problem \mathcal{A} .
 - Run the black box for \mathcal{A} on A .
 - Convert the solution for A back into a solution for B .

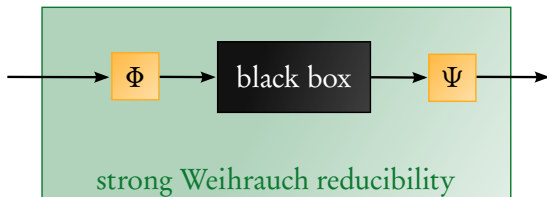
Reminder: Weihrauch reducibilities



- 1 Φ and Ψ are both Turing functionals.
- 2 **Weak reducibility.** The decoding procedure Ψ has access to the original input.
- 3 That is, if B is the black box, the computed function is

$$x \mapsto \Psi(\langle B(\Phi(x)), x \rangle).$$

Reminder: Weihrauch reducibilities



- 1 Φ and Ψ are both Turing functionals.
- 2 **Strong reducibility.** The decoding procedure Ψ *does not* have access to the original input.
- 3 That is, if B is the black box, the computed function is

$$x \mapsto \Psi(B(\Phi(x))).$$

The principles LAY and RD

1 Let \mathcal{U} be a universal ML-test.

2 **Definition (Brattka, Gherardi, Hölzl).**

$$\text{LAY}_{\mathcal{U}}: \begin{array}{l} \text{MLR} \implies \omega \\ X \longmapsto \{i: X \notin \mathcal{U}_i\}. \end{array}$$

3 **Definition.** $\text{RD}_{\mathcal{U}}: \begin{array}{l} \text{MLR} \longrightarrow \omega \\ X \longmapsto \text{rd}_{\mathcal{U}}(X). \end{array}$

(We use $\text{RD}_{\mathcal{U}}$ for the principle, and $\text{rd}_{\mathcal{U}}$ for the property of sequences.)

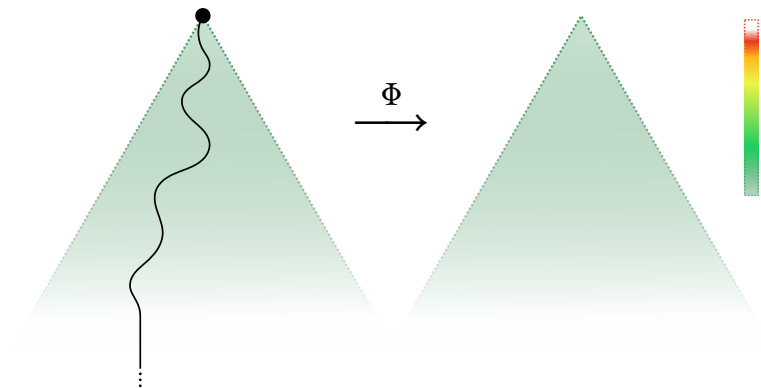
4 **Intuition.** “For a given sequence X , what is its color?”

Strong equivalence of different LAY's

1 Proposition. If \mathcal{U} and \mathcal{V} are universal ML-tests, then

$$\text{LAY}_{\mathcal{V}} \leq_{s\mathbb{W}} \text{LAY}_{\mathcal{U}}.$$

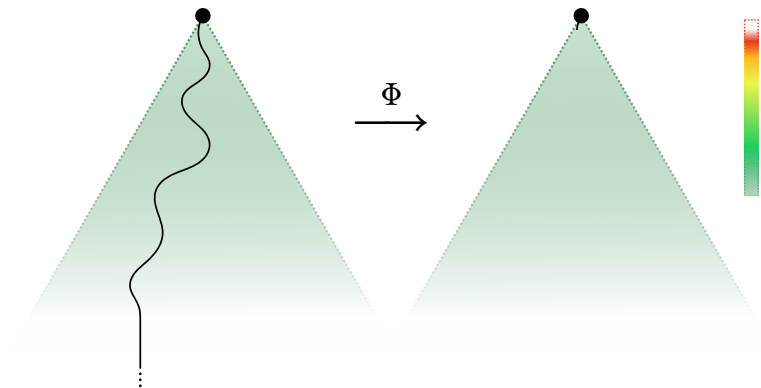
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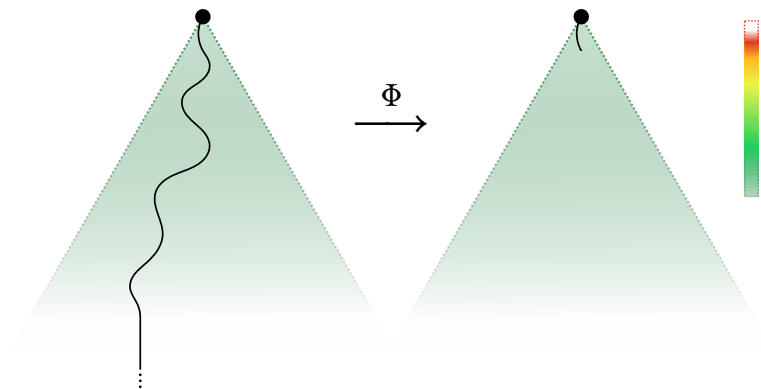
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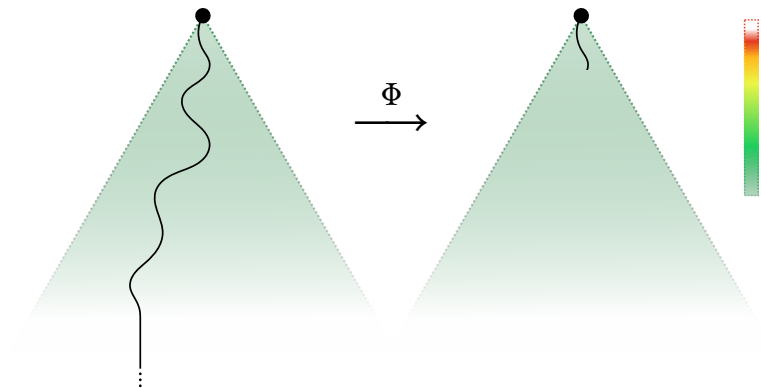
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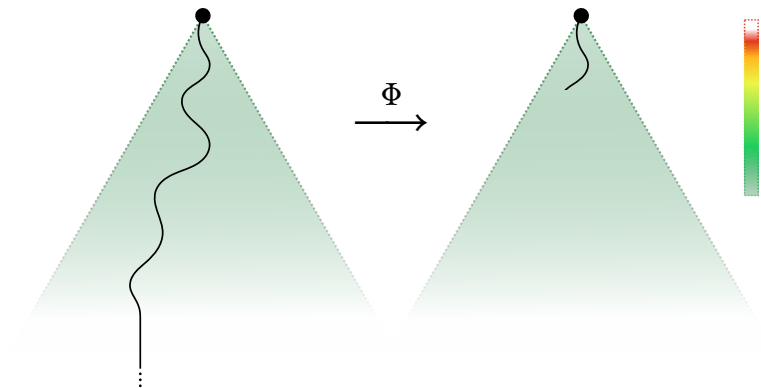
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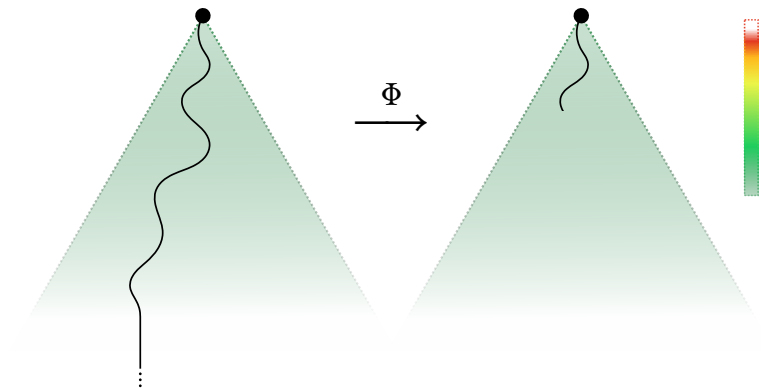
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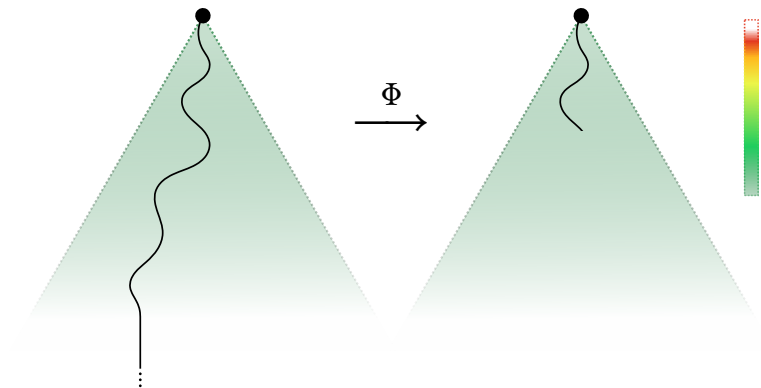
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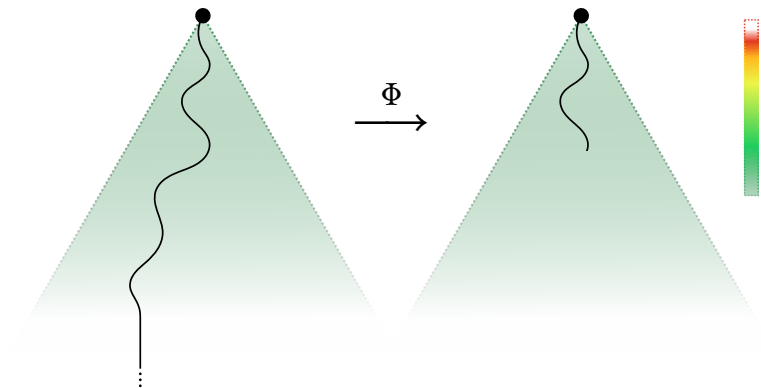
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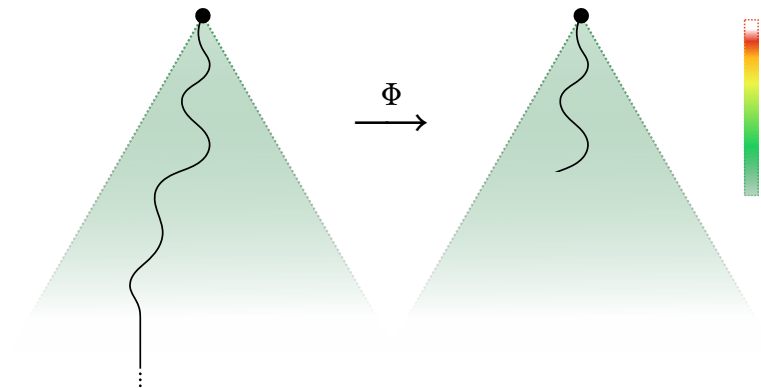
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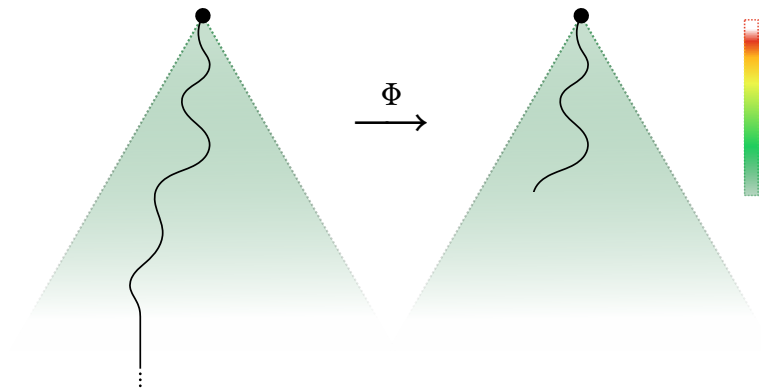
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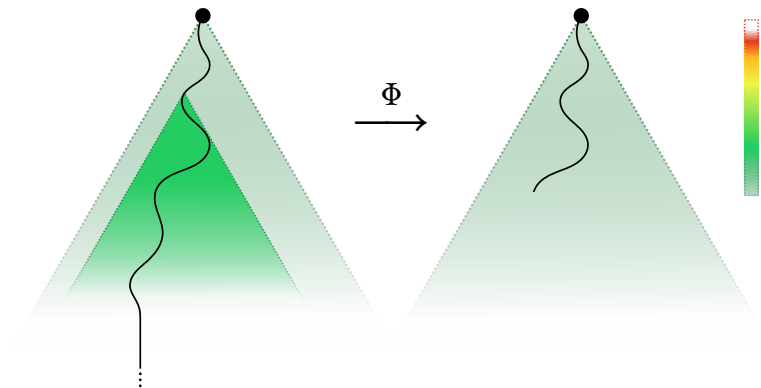
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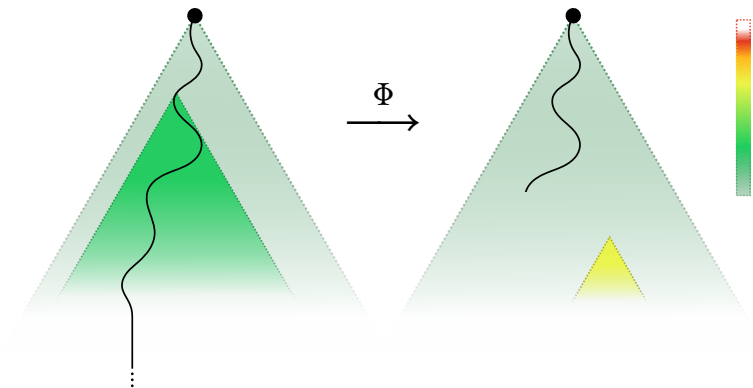
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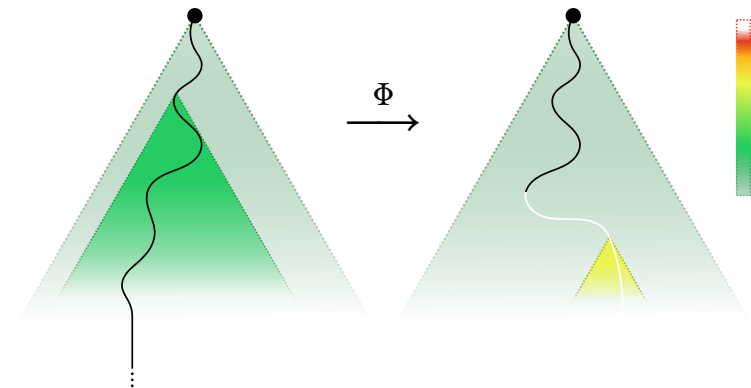
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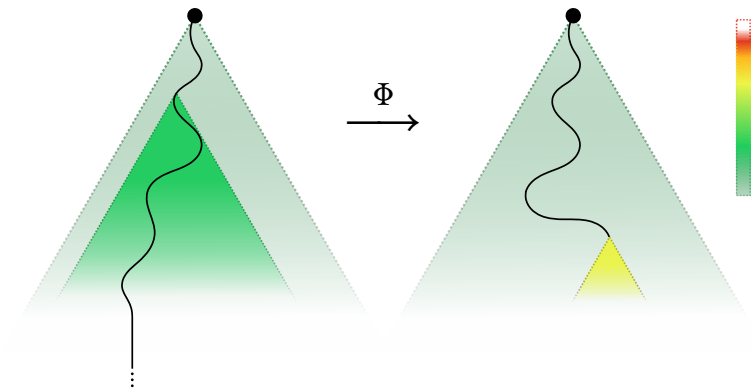
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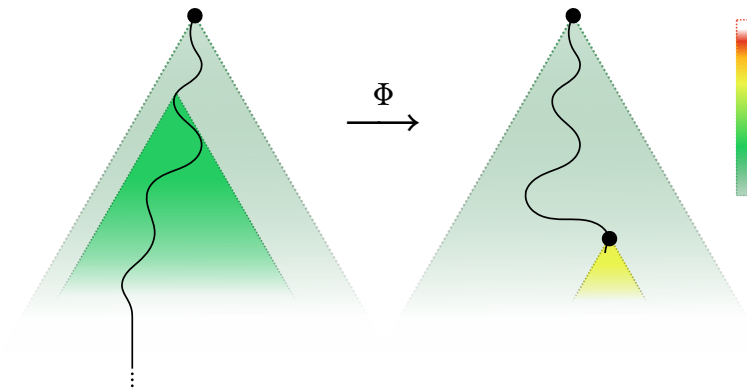
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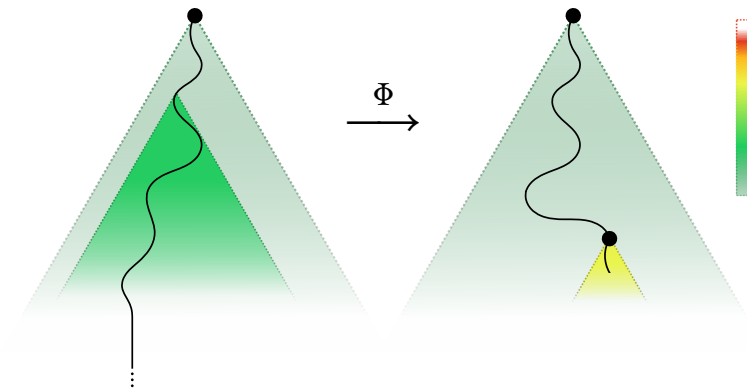
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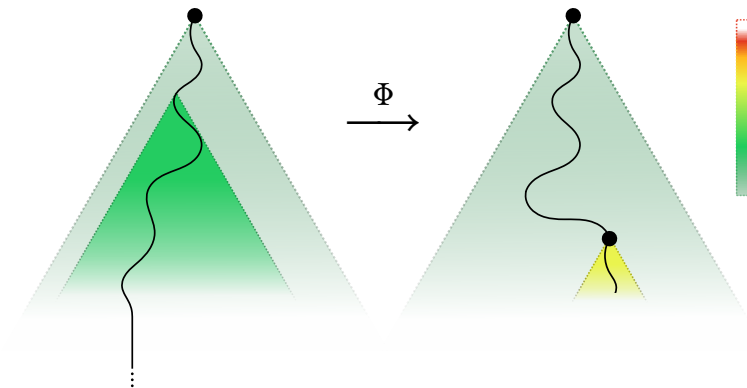
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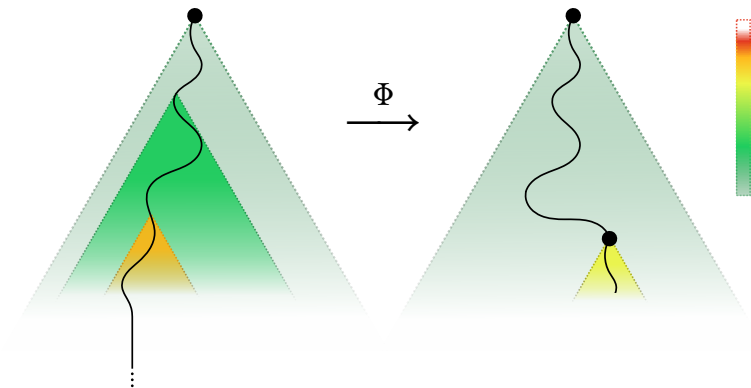
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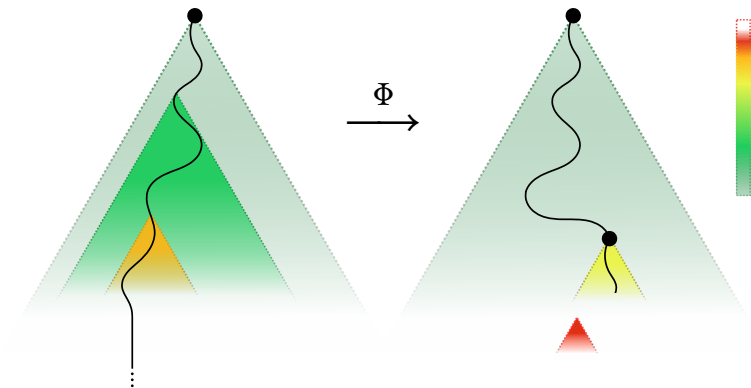
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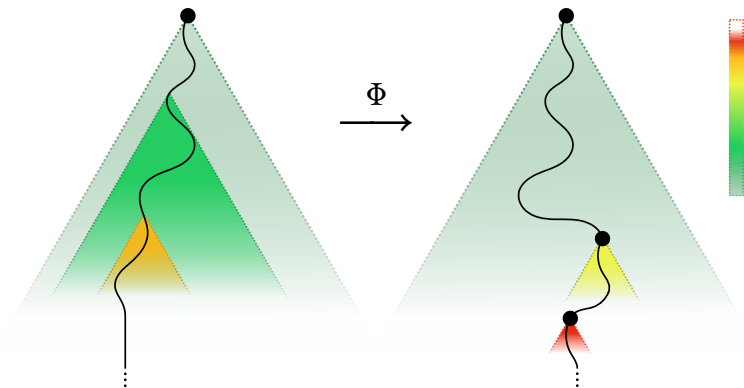
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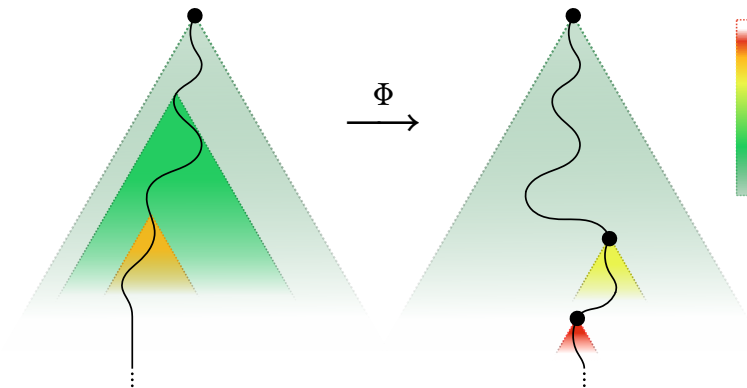
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Weak equivalence of LAY and RD

1 Theorem. Let \mathcal{U} and \mathcal{V} be universal ML-tests. Then

$$\text{RD}_{\mathcal{V}} \equiv_{\mathbb{W}} \text{LAY}_{\mathcal{U}}.$$

Proof. Same idea as for the previous proof, but more involved.

2 Proposition. Let \mathcal{U} and \mathcal{V} be universal ML-tests. Then

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- 2 Article reference.**

- Annals of Pure and Applied Logic, 2015.
- Contains further results, e.g., on the idempotence of LAY.

- 3 Related article.**

- Davie, Fouché & Pauly:
Weihrauch-completeness for layerwise computability
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- Related results and some overlap.

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Thank you for your attention.