

# Randomness for capacities

with applications to random closed sets

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# Introduction and Motivation

# The main idea

- All measures are additive:

$$\mu(A \cup B) = \mu(A) + \mu(B) \quad (A \text{ and } B \text{ disjoint})$$

- However there are also **nonadditive** “measures”, in particular
  - subadditive “measures” (a.k.a. **capacities**)

$$\mu(A \cup B) \leq \mu(A) + \mu(B) \quad (A \text{ and } B \text{ disjoint})$$

- superadditive “measures”

$$\mu(A \cup B) \geq \mu(A) + \mu(B) \quad (A \text{ and } B \text{ disjoint})$$

- Martin-Löf randomness can be extended to nonadditive “measures”.
- This provides a **unified framework** for many notions in randomness.

# Examples of randomness for capacities (subadditive)

- strong  $s$ -randomness (resp. strong  $f$ -randomness) . . . . (Reimann and many others)
  - = randomness on  $s$ -dimensional (resp.  $f$ -weighted) Hausdorff capacity
- $s$ -energy randomness . . . . . (Diamondstone/Kjos-Hanssen)
  - = randomness on  $s$ -dimensional Riesz capacity
- MLR for a class of measures . . . . . (Bienvenu/Gács/Hoyrup/Rojas/Shen)
  - = randomness on the corresponding upper envelope capacity
- members of a MLR closed set
  - MLR closed sets . . . . . (Barmpalias/Brodhead/Cenzer/Dashti/Weber)
  - zeros of MLR Brownian motion . . . . . (Kjos-Hanssen/Nerode and A/B/S)
  - image of MLR  $n$ -dim. Brownian motion . . . . . (Allen/Bienvenu/Slaman)
  - double points of MLR planar BM . . . . . (Allen/Bienvenu/Slaman)
  - = randomness on the corresponding intersection capacity
- (Unfinished work) Lebesgue points of all computable Sobolev  $W^{n,p}$  functions
  - = (some sort of) Schnorr randomness on  $n,p$ -Bessel capacity

# More examples

- Randomness for superadditive measures loosely corresponds to
  - randomness for semimeasures (studied by Levin and Bienvenu/Hölzl/Porter/Shafer)
- Randomness for capacities can be used to characterize
  - effective dimension (studied by Lutz and everyone else)

# An application: members of random closed sets

Random closed sets of Barmpalias/Brodhead/Cenzer/Dashti/Weber

- Construct a  $\{0, 1\}$ -tree with no dead-ends:
  - Branch only left with probability  $1/3$ .
  - Branch only right with probability  $1/3$ .
  - Branch both directions with probability  $1/3$ .
- A MLR tree is one constructed with a MLR in  $3^{\mathbb{N}}$ .
- A **BBCDW MLR closed set** is the set of paths through a MLR tree.
- What are the elements of a BBCDW MLR closed set?

## Theorem (Diamondstone and Kjos-Hanssen)

- *If  $z$  is MLR on some probability measure  $\mu$  satisfying the condition*

$$\iint |x - y|^{-\log_2(3/2)} d\mu(x) d\mu(y) < \infty,$$

- *then  $z$  is a member of some BBCDW MLR closed set.*

- D/K-H conjectured the converse holds. **I will show they were correct!**

# Definitions and Basic Results

# Nonadditive measures

## Definition

A **regular nonadditive measure** on  $2^{\mathbb{N}}$  or  $\mathbb{R}^n$  is a set function  $C$  such that

- $C$  is defined on all open and closed sets,
- $C$  is finite on compact sets,
- $C(\emptyset) = 0$ ,
- $C$  is monotone:  $C(A) \leq C(B)$  for  $A \subseteq B$ ,
- $C$  continuous from below on open sets:  $C(U_n) \uparrow C(U)$  if  $U_n \uparrow U$ ,
- $C$  continuous from above on compact sets:  $C(K_n) \downarrow C(K)$  if  $K_n \downarrow K$ ,
- $C$  is outer regular:  $C(A) = \inf\{C(U) : U \text{ open and } A \subseteq U\}$ ,
- $C$  is inner regular:  $C(A) = \sup\{C(K) : K \text{ compact and } K \subseteq A\}$ .

## Definition

If  $C$  is subadditive ( $\mu(A \cup B) \geq \mu(A) + \mu(B)$ ) then call  $C$  a **capacity**.

# Computable nonadditive measures

## Definition

Say that a regular nonadditive measure is **computable** if

- $C$  is uniformly lower-semicomputable on  $\Sigma_1^0$  sets, and
- $C$  is uniformly upper-semicomputable on  $\Pi_1^0$  sets.

Or equivalently on  $2^{\mathbb{N}}$ ,

- $C$  is uniformly computable on clopen sets (not just cylinder sets).

## Definition

A **Martin-Löf test**  $(U_n)$  is a uniform sequence of  $\Sigma_1^0$  sets such that  $C(U_n) \leq 2^{-n}$ .  
A point  $x$  is **Martin-Löf random** on  $C$  if  $x \notin \bigcap_n U_n$  for all ML tests.

## Remark

If  $C$  is countably subadditive,  $C(\bigcup_n A_n) \leq \sum C(A_n)$ ,

- There is a universal test (same proof!).
- Can also use Solovay tests, integrable tests, etc.

# An important lemma

For two capacities, write

$$C_1 =^{\times} C_2$$

if they are equal up to a multiplicative constant, i.e. there are two constants  $0 < a < b$  such that for all  $A$ ,

$$aC_1(A) \leq C_2(A) \leq bC_1(A).$$

## Lemma

For two computable capacities, if  $C_1 =^{\times} C_2$  then  $C_1$  and  $C_2$  have the same ML randoms.

# Randomness for non-computable measures

- It is well known that MLR can be extended to noncomputable probability measures (Levin, Reimann/Slaman, Day/Miller, B/G/H/R/S)

## Definition

A point  $x \in X$  is **Martin-Löf random on  $\mu$**  if  $x$  is not covered by any ML tests “computable from  $\mu$ ”.

- Let  $\mathcal{M}^+(X)$  denote the space of all finite Borel measures on  $X$ .

## Lemma

If  $x$  is random on  $\mu \in \mathcal{M}^+(X)$ , then  $x$  is random on the prob. measure  $\mu/\mu(X)$ .

- Randomness can also be developed for non-computable capacities, but this is not needed in this talk.

# Effective compactness

On any computable metric space  $X$  (e.g.  $2^{\mathbb{N}}$ ,  $\mathbb{R}^n$ ,  $\mathcal{M}^+(2^{\mathbb{N}})$ ,  $\mathcal{M}^+(\mathbb{R}^n)$ ) let  $\mathcal{K}(X)$  denote the space of nonempty compact sets.

## Definition

A nonempty compact set  $K$  is said to be<sup>1</sup> **computable in  $\mathcal{K}(X)$**  iff any of the following equivalent conditions hold:

- 1  $K$  is computable in the Hausdorff metric on  $\mathcal{K}(X)$ .
- 2 The distance function  $x \mapsto \text{dist}(x, K)$  is computable.
- 3  $K$  is the image of a total computable map  $2^{\mathbb{N}} \rightarrow X$ .
- 4  $K$  is  $\Pi_1^0$  and contains a computable dense subsequence.
- 5 (On  $2^{\mathbb{N}}$ )  $K$  is the set of paths of a computable tree with no dead branches.
- 6 (On  $\mathbb{R}$ ) Both  $\{a, b \in \mathbb{Q} : K \cap (a, b) \neq \emptyset\}$  and  $\{a, b \in \mathbb{Q} : K \cap [a, b] = \emptyset\}$  are  $\Sigma_1^0$ .
- 7 The map  $f \mapsto \max(K, f)$  is uniformly computable for continuous  $f$ .

<sup>1</sup>Others use the terms “located set”, “computable set”, or “decidable set”. The term “effectively compact set” is usually reserved for a weaker notion.

# Some Examples of Capacities and Characterizations of their Randoms Points

# Upper envelopes / Randomness for classes of measures

- Let  $\mathcal{C}$  be a compact subset of  $\mathcal{M}^+(X)$  (finite Borel measures on  $X$ ).
- The **upper envelope capacity** is defined as

$$\text{Cap}_{\mathcal{C}}(A) = \sup_{\mu \in \mathcal{C}} \mu(A).$$

## Theorem

If  $\mathcal{C}$  is computable in  $\mathcal{K}(\mathcal{M}^+(X))$ , then  $\text{Cap}_{\mathcal{C}}$  is a computable capacity.

## Theorem (Basically Bienvenu-Hoyrup-Gács-Rojas-Shen)

If  $\mathcal{C}$  is computable in  $\mathcal{K}(\mathcal{M}^+(X))$ , the following are equivalent.

- $x$  is random on the upper envelope  $\text{Cap}_{\mathcal{C}}$ .
- $x$  is random on some measure  $\mu \in \mathcal{C}$ .

# Riesz capacity / Energy randomness

- $s$ -energy of a measure  $\mu \in \mathcal{M}^+(\mathbb{R}^n)$

$$\text{Energy}_s(\mu) := \begin{cases} \iint |x-y|^{-s} d\mu(x)d\mu(y) & 0 < s \leq n \\ \iint \log \frac{1}{|x-y|} d\mu(x)d\mu(y) & s = 0 \end{cases}$$

- $s$ -Riesz capacity (version 1)

$$\mathbf{C}_s(A) := \sup \{ \mu(A) \mid \mu \in \mathcal{M}^+(\mathbb{R}^n), \text{Energy}_s(\mu) \leq 1 \}$$

## Theorem

For computable  $s$ ,

- $\{ \mu \mid \mu \in \mathcal{M}^+(\mathbb{R}^n), \text{Energy}_s(\mu) \leq 1 \}$  is a computable set in  $\mathcal{K}(\mathcal{M}^+(\mathbb{R}^n))$ .
  - $\mathbf{C}_s$  is a computable (upper envelope) capacity
  - The following are equivalent:
    - $x$  is random on the Riesz capacity  $\mathbf{C}_s$ .
    - $x$  is random on some measure  $\mu \in \mathcal{M}^+(\mathbb{R}^n)$  such that  $\text{Energy}_s(\mu) \leq 1$ .
    - $x$  is random on some probability measure  $\mu$  such that  $\text{Energy}_s(\mu) < \infty$
- $\Leftrightarrow$ :  ***$s$ -energy random.***

# Riesz capacity / Energy randomness

- $s$ -Riesz capacity (version 1)

$$C_s(A) := \sup \{ \mu(A) \mid \mu \in \mathcal{M}^+(\mathbb{R}^n), \text{Energy}_s(\mu) \leq 1 \}.$$

- $s$ -Riesz capacity (version 2)

$$\text{Cap}_s(A) := \sup \left\{ \frac{1}{\text{Energy}_s(\mu)} \mid \mu \in \mathcal{M}_1(A) \right\}.$$

## Theorem

$$\text{Cap}_s = (C_s)^2.$$

## Corollary

The following are equivalent

- $x$  is random on the Riesz capacity  $\text{Cap}_s$ .
- $x$  is random on the Riesz capacity  $C_s$ .
- $x$  is  $s$ -energy random.

## Random compact sets / Members of MLR compact sets

- For a probabilist, a **random set** is a process which randomly generates a set, i.e. a set-valued random variable  $Z$ .
- Each random compact set corresponds to a probability measure  $\mathbf{P}_Z$  on the space of compact sets  $\mathcal{K}(2^{\mathbb{N}})$ .
- A compact set  $K$  is a **MLR compact set** if it is MLR on  $\mathbf{P}_Z$ .
- The **intersection capacity** is defined as follows:

$$\mathbb{T}_Z(A) = \mathbf{P}\{Z \cap A \neq \emptyset\} = \mathbf{P}_Z \left\{ K \in \mathcal{K}(2^{\mathbb{N}}) : K \cap A \neq \emptyset \right\}.$$

- $\mathbb{T}_Z(A)$  is the probability that  $A$  intersects some random compact set.

### Theorem (R.)

For computable measure  $\mathbf{P}_Z$  on  $\mathcal{K}(2^{\mathbb{N}})$ , the following are equivalent.

- $x$  is a random for the intersection capacity  $\mathbb{T}_Z$ .
- $x$  is a member of some  $\mathbf{P}_Z$ -MLR compact set.

# Applications of Potential Theory to Random Compact Sets

# Proof of Diamondstone and Kjos-Hanssen's conjecture

## Conjecture (Diamondstone/Kjos-Hanssen)

The following are equivalent.

- $x$  is a member of some BBCDW MLR compact set.
- $x$  is  $\log_2(3/2)$ -energy random.
  
- Let  $T_{BBCDW}$  be the intersection capacity of the BBCDW random set.
- Recall  $\text{Cap}_{\log_2(3/2)}$  denotes the  $\log_2(3/2)$ -Riesz capacity.

## Equivalent Conjecture

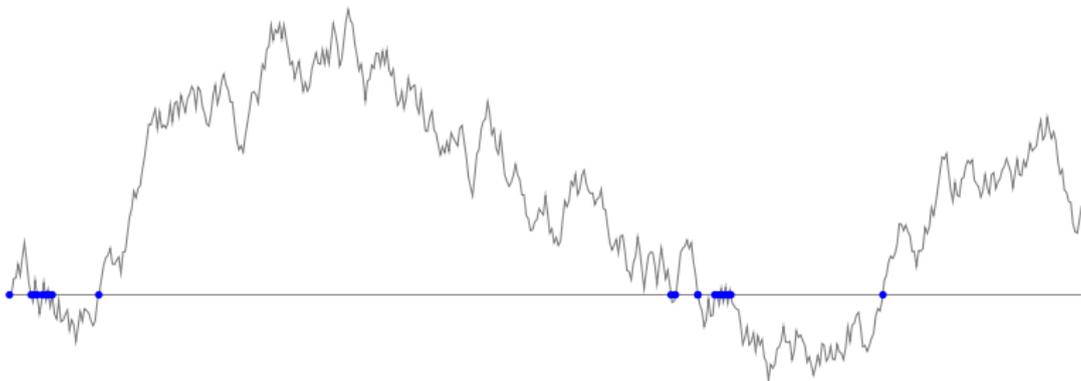
$T_{BBCDW}$  and  $\text{Cap}_{\log_2(3/2)}$  have the same randoms.

The conjecture follows from...

## Theorem (Lyons)

$$T_{BBCDW} =^{\times} \text{Cap}_{\log_2(3/2)}.$$

# Zeros of Brownian motion



The zeros of Brownian motion  $B$  form a random closed set.

$$Z_B = \{t \in [a, b] : B(t) = 0\}$$

Question (Allen/Bienvenu/Slaman)

Which reals are zeros of some MLR Brownian motion?

# Zeros of a MLR Brownian motion

Let  $T_{Z_B}$  be the intersection capacity of the random closed set  $Z_B$  of zeros.

$$T_{Z_B}(A) := \mathbb{P}\{Z_B \cap A \neq \emptyset\}$$

## Theorem (Kakutani)

$T_{Z_B} = {}^\times \text{Cap}_{1/2}$  when restricted to an interval  $[a, b]$  ( $0 < a < b$ ).

## Corollary

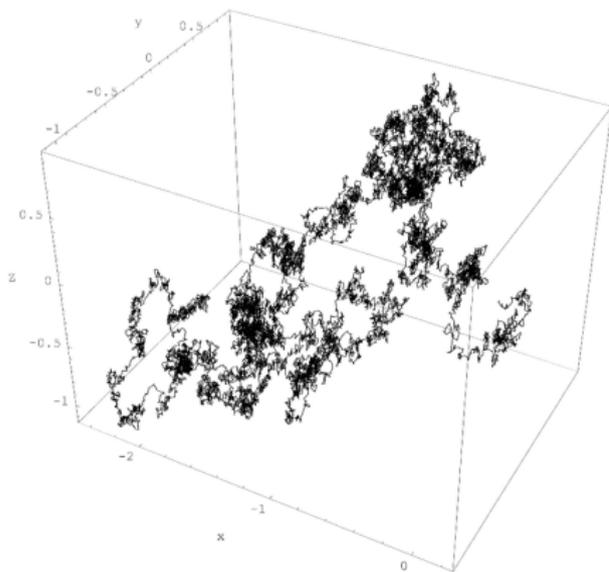
$T_{Z_B}$  and  $\text{Cap}_{1/2}$  have the same randoms (in  $[a, b]$ )

## Theorem (R.—Allen/Bienvenu/Slaman were very close!)

*The following are equivalent.*

- *$x$  is a zero time of some MLR Brownian motion.*
- *$x$  is 1/2-energy random.*

# Image of an $n$ -dimensional Brownian motion



The image of an  $n$ -dimensional Brownian motion forms a random closed set.

$$B([a, b]) = \{B(t) : a \leq t \leq b\}$$

**Question (Allen/Bienvenu/Slaman)**

Which points are in the image of some  $n$ -dimensional MLR Brownian motion?

**Figure :** Image of a 3D Brownian motion ([http://en.wikipedia.org/wiki/Wiener\\_process](http://en.wikipedia.org/wiki/Wiener_process))

# Image of a MLR $n$ -dimensional Brownian motion

Let  $\tau$  be the intersection capacity of  $B([a, b])$ .

$$\tau_{B([a, b])}(A) := \mathbb{P}\{B([a, b]) \cap A \neq \emptyset\}$$

## Theorem (Kakutani)

$\tau_{B([a, b])} =^{\times} \text{Cap}_{n-2}$  when  $n \geq 2$ .

## Corollary

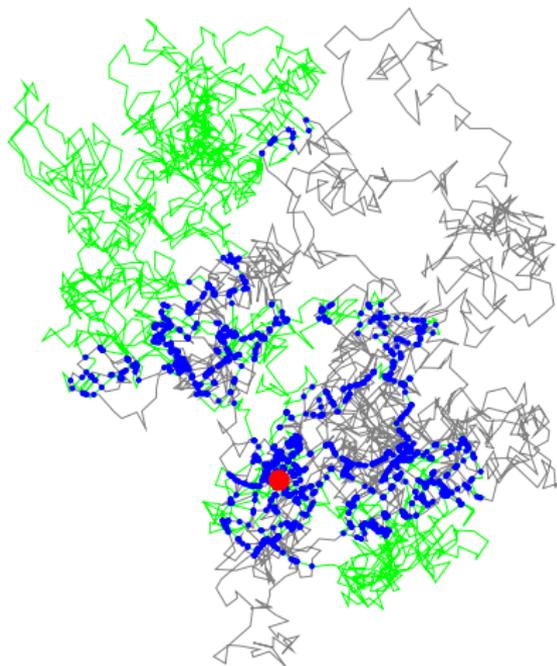
$\tau_{B([a, b])}$  and  $\text{Cap}_{n-2}$  have the same randoms (for  $n \geq 2$ ).

## Theorem (R.)

The following are equivalent for  $x \in \mathbb{R}^n$  ( $x \neq 0$ ) and  $n \geq 2$ .

- $x$  is in the image of some  $n$ -dimensional MLR Brownian motion.
- $x$  is  $(n-2)$ -energy random.

# Double points of planar Brownian motion



- The double points of a planar Brownian motion are the points where it intersects itself.

## Question (Allen/Bienvenu/Slaman)

Which points are the double points of some MLR planar Brownian motion?

- Closely related, consider the intersection points of two independent planar Brownian motions which start at the origin.
- This forms a random closed set.

**Figure :** Intersection (in blue) of two planar Brownian motions starting at the origin.

# Double points of MLR planar BM

- Let  $\mathsf{T}_{B_1 \cap B_2}$  be the intersection capacity of random closed set  $B_1([a, b]) \cap B_2([a, b])$ .

$$\mathsf{T}_{B_1 \cap B_2}(A) := \mathbb{P}\{B_1([a, b]) \cap B_2([a, b]) \cap A \neq \emptyset\}$$

- Define  $\text{Cap}_{\log^2}$  and **log<sup>2</sup>-energy random** using the energy

$$\iint (\log|x - y|)^2 d\mu(x) d\mu(y).$$

Theorem (Fitzsimmons/Salisbury, also Permantal/Peres)

$$\mathsf{T}_{B_1 \cap B_2} =^{\times} \text{Cap}_{\log^2}.$$

Theorem (R.)

The following are equivalent for  $x \in \mathbb{R}^2$  ( $x \neq 0$ ).

- $x$  is a double point of some MLR planar Brownian motion.
- $x$  is in the intersection of some pair of independent MLR planar BM.
- $x$  is log<sup>2</sup>-energy random.

# Capacities and Effective Hausdorff Dimension

# Hausdorff capacity vs Hausdorff measure

- **s-Hausdorff capacity** (a.k.a. **s-Hausdorff content**) of radius  $r \in (0, \infty]$ .

$$\mathcal{H}_r^s(A) := \inf \sum_i 2^{-s|\sigma_i|}$$

where the infimum is over covers  $\bigcup_i [\sigma_i] \supseteq A$  such that  $2^{|\sigma_i|} \leq r$ .

- s-dimensional **Hausdorff measure**

$$\mathcal{H}^s(A) := \lim_{r \rightarrow 0} \mathcal{H}_r^s(A)$$

- Same null sets:

$$\mathcal{H}^s(A) = 0 \quad \Leftrightarrow \quad \mathcal{H}_r^s(A) = 0.$$

- $\mathcal{H}^s$  is an ugly measure (open sets have infinite measure!)
- $\mathcal{H}_r^s$  is a nice capacity.
- $\mathcal{H}_\infty^s$  is called **vehement weight** by some in algorithmic randomness

# Frostman's Lemma / Strong $s$ -randomness

## Frostman's Lemma

The following are equivalent.

- $\mathcal{H}_\infty^s(A) > 0$
- $\mu(A) > 0$  for some prob. measure  $\mu$  such that  $\mu(\sigma) \leq 2^{-s|\sigma|}$  for all  $\sigma$ .
- A point  $x \in 2^\mathbb{N}$  is **strongly  $s$ -random** if  $KM(x \upharpoonright n) \geq sn + O(1)$ .  
( $KM$  is a priori complexity.)
- A point  $x \in 2^\mathbb{N}$  is **vehemently  $s$ -random** if  $x$  is  $\mathcal{H}_\infty^s$ -random.
- A point  $x \in 2^\mathbb{N}$  is  **$s$ -capacitable** if  $x$  is  $\mu$ -random for some prob. measure  $\mu$  s.t.  $\mu(\sigma) \leq 2^{-s|\sigma|}$  for all  $\sigma$ .

## Effective Frostman's Lemma (Reimann, Kjos-Hanssen)

The following are equivalent.

- $x$  is strong  $s$ -random.
- $x$  is  $s$ -vehemently random ( $\mathcal{H}_\infty^s$ -random).
- $x$  is  $s$ -capacitable.

# Effective dimension (Frostman's Theorem)

## Theorem (Frostman's Theorem)

$$\dim(A) = \sup\{s \mid \mathcal{H}_r^s(A) > 0\} = \sup\{s \mid \text{Cap}_s(A) > 0\}$$

## Theorem (Effective Frostman's Theorem (Reimann, Diamondstone/K-H))

$$\text{cdim}(x) = \sup\{s \mid x \text{ is strongly } s\text{-random}\} = \sup\{s \mid x \text{ is } s\text{-energy random}\}$$

## Question (Reimann)

Are strong  $s$ -randomness and  $s$ -energy randomness the same?

## Answer

No.

# Strong $s$ -randomness vs. $s$ -energy randomness

## Question

Are strong  $s$ -randomness and  $s$ -energy randomness equal?

Theorem (Maz'ya/Khavin, also in Adams/Hedberg book)

*There is a(n effective) closed set  $E$  such that  $\mathcal{H}_r^s(E) > 0$ , but  $\mathbf{Cap}_s(E) = 0$ .*

Corollary (R.)

*There is an  $x$  which is strongly  $s$ -random, but not  $s$ -energy random.*

# Summary of known results

## Known Results (No arrows reverse)

$\text{cdim } x > s$

$\Rightarrow x$  is  $s$ -energy random  $\Leftrightarrow x$  is random on  $\text{Cap}_s$

$\Rightarrow x$  is strongly  $s$ -random  $\Leftrightarrow x$  is random on  $\mathcal{H}_r^s$

$\Rightarrow x$  is weakly  $s$ -random (that is  $K(x \upharpoonright n) \geq sn + O(1)$ )

$\Rightarrow \text{cdim } x \geq s$ .

## Remark

Weak  $s$ -randomness is not associated (at least in any nice way) with a capacity.

# Cylinder sets do not determine randomness

- Let  $T_{BBCDW}$  be the intersection capacity of the BBCDW random sets.
- Let  $\mathcal{H}_\infty^{\log_2(3/2)}$  be the Hausdorff capacity for dimension  $s = \log_2(3/2)$ .
- It turns out that
  - $T_{BBCDW}([\sigma]) = \mathcal{H}_\infty^{\log_2(3/2)}([\sigma]) = (2/3)^{|\sigma|}$ .
  - $T_{BBCDW}$ -random ( $\log_2(3/2)$ -energy random)  $\neq \mathcal{H}_\infty^{\log_2(3/2)}$ -random (strong  $\log_2(3/2)$ -random).

## Remark

Therefore, the values of a capacity **on cylinder sets** is not enough to determine its randomness.

# Superadditive Measures and Semi-measures

# Semi-measures

- Semi-measures:

$$\rho(\sigma 0) + \rho(\sigma 1) \leq \rho(\sigma).$$

- Levin and Bienvenu/Hölzl/Porter/Shafer have looked at randomness for computable and left c.e. semimeasures.
- It is messy in the left c.e. case!
- Can extend  $\rho$  to the smallest superadditive measure  $C \geq \rho$ .
- **Blind (Hippocratic) randomness** is ML randomness (for non-computable measures) which doesn't use the measure as an oracle.

## Theorem (R.)

*If  $\rho$  is computable (resp. left c.e.), the following are equivalent.*

- *$x$  is random (resp. blind random) on  $C$ .*
- *$x$  is random on  $\rho$  using Bienvenu/Hölzl/Porter/Shafer definitions.*
- *$x$  is blind  $\mu$ -random for every probability measure  $\mu \geq \rho$ .*

There is so much more to say...

...but I have run out of time.

# Summary

Randomness for nonadditive measures...

- provides a unified framework for concepts in randomness
- helps us to prove new results
- simplifies the proofs of old results
- gives us 60-years-worth of theorems in capacity theory to draw from!

# Closing Thoughts

# Thank You!

These slides will be available on my webpage:

<http://www.personal.psu.edu/jmr71/>

Or just Google™ me, “Jason Rute”.