

Dimension of SLE curves.

$$\text{Hdim } K := \inf_{\delta} \inf_{\epsilon > 0} \exists \{B(x_j, \delta_j)\} : \sum \delta_j^d < \epsilon, \cup B(x_j, \delta_j) \supset K$$

↓
Hausdorff dimension.

Theorem (Beffara). For $K \leq 8$,
 $\text{Hdim}(SLE_K^-) = 1 + \frac{K}{8}$ a.s.

We will prove an upper bound.

$$\text{Let } N(\delta, K) := \min \# \{B(x_j, \delta) : K \subset \cup B(x_j, \delta)\}$$

Upper box (Minkowski) dimension:

$$\overline{\text{Mdim}}(K) := \limsup_{\delta \rightarrow 0} \frac{\log N(\delta, K)}{\log \frac{1}{\delta}}$$

Lower box (Minkowski) dimension:

$$\underline{\text{Mdim}}(K) := \liminf_{\delta \rightarrow 0} \frac{\log N(\delta, K)}{\log \frac{1}{\delta}}$$

Observe: $\text{Hdim } K \leq \underline{\text{Mdim}}(K) \leq \overline{\text{Mdim}}(K)$

Indeed, if $d > \underline{\text{Mdim}}(K) \exists \delta_j \downarrow 0 : N(\delta_j, K) < \left(\frac{1}{\delta_j}\right)^{d-\epsilon} \Rightarrow$
 $\sum \delta_j^d = \delta_j^{d-\epsilon} \delta_j^\epsilon \rightarrow 0$ as $\delta_j \rightarrow 0$. So $d \geq \text{Hdim } K$

Lemma (an easy upper bound) Let K be a random set and $\lim_{\delta \rightarrow 0} \frac{\log E(N(\delta, K))}{\log \frac{1}{\delta}} = d_1$ ($d_1 > 0$). Then $\overline{\text{Mdim}} K \leq d_1$

Pf. Take $d_2 > d_1$. Then for small δ ,
 $E(N(K, \delta)) \leq \delta^{-d_2}$. So $P(N(K, \delta) \geq \delta^{-d_2}) \leq \delta^{d_1 - d_2}$

Now take $\delta_n = 2^{-n}$, observe that

$$\limsup_{\delta \rightarrow 0} \frac{\log E(N(\delta, K))}{\log \frac{1}{\delta}} = \lim_{n \rightarrow \infty} \frac{\log E(N(2^{-n}, K))}{n \log 2}, \text{ and}$$

notice that $\sum P(N(K, 2^{-n}) \geq 2^{-nd_1}) \leq \sum 2^{n(d_1 - d_2)} < \infty$

So, by Borel-Cantelli, $N(K, 2^{-n}) \leq 2^{-nd_1}$ for large n

Proof that $\lim_{\delta \rightarrow 0} \frac{\log E(N(\delta, K) \cap A)}{\log \frac{1}{\delta}} = 1 + \frac{K}{8}, K \leq 8$
 for any open A with $\overline{A} \subset \mathbb{H}$ bounded

Take $\delta = \text{dist}(A, \mathbb{R})$.

Let $\mathcal{M} = \{z \in \mathbb{R}^2 : \text{dist}(z, A) \leq \delta\}$. $m := \#\mathcal{M} \approx \frac{1}{\delta^2}$.

Observe $N(\delta, \gamma(0, \infty) \cap A) \leq \#\{z \in \mu: \text{dist}(z, \gamma(0, \infty)) < \delta\}$

and $\#\{z \in \mu: \text{dist}(z, \gamma(0, \infty)) < \delta\} \leq N(\delta, \gamma(0, \infty) \cap A)$

But $E(\#\{z \in \mu: \text{dist}(z, \gamma(0, \infty)) < \delta\}) = \sum_{z \in \mu} P(\text{dist}(z, \gamma(0, \infty)) < \delta)$

By scaling and boundedness of A ,

$$P(\text{dist}(z, \gamma(0, \infty)) < \delta) \approx \sup_{|x| \leq M} P(\text{dist}(x+i, \gamma(0, \infty)) < C\delta)$$

for some M, C .

So, to establish the bound, we need $\exists c$.

$$c^{-1} \delta^{1-\frac{\kappa}{8}} \leq P(\text{dist}(x+i, \gamma(0, \infty)) < \delta) \leq c \delta^{1-\frac{\kappa}{8}}, \quad |x| \leq M. \quad (*)$$

Reminder. $\text{dist}(x+i, \gamma(0, \infty)) \approx \exp(-D(x))$,

$$D(x) = \lim_{t \rightarrow \infty} \frac{\ln |h_t'(x+i)|}{\Gamma_n h_t(x+i)}$$

$$(*) \quad E(e^{ibDW}) = \frac{\Gamma(\frac{2}{\kappa} + \sqrt{(\frac{2}{\kappa} - \frac{1}{2})^2 - i\frac{2b}{\kappa}}) \Gamma(\frac{2}{\kappa} - \sqrt{(\frac{2}{\kappa} - \frac{1}{2})^2 - i\frac{2b}{\kappa}})}{\Gamma(\frac{1}{2}) \Gamma(\frac{\kappa}{2} - \frac{1}{2})} F(\frac{1}{2}, \frac{\kappa}{2} - \frac{1}{2}, \frac{\kappa}{2}, x).$$

We'll need the standard estimate:

Lemma. X - random variable, $\varphi(b) = E(\exp(ibX))$ - Characteristic function

Assume that for some $u, \lambda, \varepsilon > 0$:

$$\varphi(b) = \frac{u\lambda}{\lambda - ib} + V(b),$$

V - analytic on $\{ |z| < \lambda + 2\varepsilon \}$.

$$\text{Then } P(X \geq x) = u e^{-\lambda x} + o(e^{-(\lambda + \varepsilon)x}) \quad (x \rightarrow \infty)$$

Proof. Let μ be distribution of X .

$$\text{Let } \mu_1 = u\lambda e^{-\lambda x} \mathbb{1}_{x > 0} dx$$

$$\left(\text{and } \int e^{i\lambda b} d\mu_1(b) = \frac{u\lambda}{\lambda - ib} \right).$$

$$\mu_2 = \mu - \mu_1. \quad \text{Then } \int e^{i\lambda b} d\mu_2(b) = V(b).$$

Since V is analytic in $\{ |z| < \lambda + 2\varepsilon \}$,

$$\mu_2(\mathbb{R}^+) = o(e^{-(\lambda + \varepsilon)x}) \quad \text{inverse Fourier!} \quad \text{So}$$

$$\mu_2([X, \infty) = \bar{\sigma}(e^{-(\lambda+\varepsilon)x}) \quad \text{inverse ...}$$

$$P(X \geq x) = \mu_1([X, \infty) + \mu_2([X, \infty)) = ue^{-\lambda x} + \bar{\sigma}(e^{-(\lambda+\varepsilon)x})$$

From (*), we can use Lemma for $\lambda = 1 - \frac{k}{8}$,
 $\varepsilon = \frac{z}{k} + \frac{k}{8} - 1$.

$$\text{Indeed, } \Gamma(z) = \sum_{k=0}^{\infty} \frac{(-1)^k}{k!} \frac{1}{z+k} + \int_0^{\infty} e^{-t} t^{z-1} dt$$

$$\Gamma\left(\frac{z}{k} - z\right) \Gamma\left(\frac{z}{k} + z\right) = \frac{\gamma_k}{\frac{\gamma}{k^2} - z^2} + H(z),$$

where $H(z)$ is analytic in $\{|z| < \frac{z}{k} + 1 + \varepsilon\}$

Plug in (*)