

First, extend the definitions of  $g_t$  and  $f_t$  for  $t < 0$ : reverse SLE flow.

Lemma. ( $t$  and  $-t$  change roles of  $f$  and  $g$ .)

$z \mapsto g_{-t}(z)$  has the same distribution as

$$z \mapsto \widehat{f}_t(z) - \beta(\kappa t)$$

Proof Fix  $t_1 \in \mathbb{R}$ . Let  $\xi(t) := \beta(\kappa t)$   
 $\tilde{\xi}(t) := \xi(t_1 + t) - \xi(t_1) \stackrel{\text{Law}}{\sim} \xi(t)$

$$\tilde{g}_t(z) := g_{t_1+t} \circ f_{t_1}(z + \xi(t_1)) - \tilde{\xi}(t)$$

$$\tilde{g}_0(z) = z, \quad \tilde{g}_{-t_1}(z) = \widehat{f}_{t_1}(z) - \xi(t_1)$$

$$\text{So } \partial_t \tilde{g}_t(z) = \frac{2}{\tilde{g}_t(z) + \tilde{\xi}(t) - \tilde{\xi}(t_1)} = \frac{2}{\tilde{g}_t(z) - \tilde{\xi}(t)} - \text{SLE}_\kappa.$$

Take  $t = -t_1$ .

Remark.  $\text{Im } g_t(z)$  is decreasing in  $t \forall z \in \mathbb{H}$ ,  
 $t < T(z) = \sup \{ t : z \in f_t(\mathbb{H}) \}$ .

Time-change:  $T_u \sim T_u(z) : \log \text{Im } g_{T_u}(z) = u$

Well-defined: Need to check that  $T_u < \infty$ , i.e.

$$\lim_{t \rightarrow T(z)} \text{Im } g_t(z) = 0,$$

$$\lim_{t \rightarrow -\infty} \text{Im } g_t(z) = \infty$$

Proof (of well-defined):

$$|\partial_t g_t(z)| \approx 2 |g_t(z) - \xi(t)|^{-1}$$

$$\text{Let } \bar{\xi}(t) = \sup_{0 \leq s \leq t} |\xi(s)| = \sup_{0 \leq s \leq t} |\beta(\kappa s)|.$$

Observe: if  $|g_s(z)| > \bar{\xi}(t)$ , then

$$|\partial_s g_s(z)| \leq \frac{2}{|g_s(z)| - \bar{\xi}(t)}$$

Let  $U := \{s \leq t : |g_s(z)| > \bar{\zeta}(t)\}$ .

$$a_s := |g_s(z)| - \bar{\zeta}(t). \quad \text{On } U, \quad (a_s^2)' = 2 a_s a_s' \leq 4$$

By integrating over components of  $U$ ,  
we get  $|g_t(z)| = a_t + \bar{\zeta}(t) \leq \bar{\zeta}(t) + 2\sqrt{t+1} + |z|$ .

Let now  $y_t := \text{Im } g_t(z)$ . Then  $-\frac{\partial}{\partial t} \log y_t = \frac{\text{Im } \frac{2}{g_t(z) - \bar{\zeta}(t)}}{\text{Im } g_t(z)} \geq \frac{2}{(t+1)\bar{\zeta}(t) + 2\sqrt{t+1}}$

$$\bar{\zeta}(t) = \max_{|s| \leq t} |B(\kappa_s)|$$

By Khinchin's law of iterated logarithm,

$$\bar{\zeta}(t) = \bar{o}(\sqrt{t}).$$

So  $\frac{\partial}{\partial t} \log y_t \sim \frac{1}{\sqrt{t}}$  as  $t \rightarrow \pm\infty$ , so it is not integrable over  $(-\infty, 0]$  or  $[0, +\infty)$   $\blacksquare$

$$u(t) := \log \text{Im } g_t(z)$$

$$\partial_t u = \frac{-2}{|g_t(z) - \bar{\zeta}(t)|^2}$$

inverse time change, time  $u$  is  $z$ -dependent.

Fix  $\hat{z} = \hat{x} + i\hat{y} \in \mathbb{H}$ .  $\forall u \in \mathbb{R}$ , let  $z(u) := g_{T_u(\hat{z})}(\hat{z}) - \bar{\zeta}(T_u(\hat{z}))$

$$x(u) := \text{Re } z(u)$$

$$y(u) = \text{Im } z(u) = e^u$$

$$\psi(u) := \frac{\hat{y}}{y(u)} |g'_{T_u}(\hat{z})| = \hat{y} e^{-u} |g'_{T_u}(\hat{z})|$$

**Theorem.** Let  $\nu = -\text{sign}(\log \hat{y})$  (i.e.  $\nu = 1 \Leftrightarrow \hat{y} < 1$ )

$$b \in \mathbb{R}, \quad a = 2b + \nu \kappa \frac{b(1-b)}{2}, \quad \lambda = 4b + \nu \kappa \frac{b(1-2b)}{2}$$

$$\text{Set } F(\hat{z}) = F_b(\hat{z}) := \hat{y}^a E \left( (1+x(t))^2 |g'_{T_t}(\hat{z})|^a \right)$$

$$\text{Then } F(\hat{z}) = \left( 1 + \left( \frac{\hat{x}}{\hat{y}} \right)^2 \right)^b \hat{y}^\lambda$$

Proof. Note that

$$du = \frac{-2}{|g_t(z) - \bar{\zeta}(t)|^2} dt = -2 |z(u)|^{-2} dt$$

Define a new process:

$$\widehat{B}(u) := -\sqrt{\frac{2}{\kappa}} \int_0^u |z(u,t)|^{-1} d\beta(t) \quad \widehat{B} - \text{local martingale}$$

$$d\langle \widehat{B}, \widehat{B} \rangle_t = -2|z|^{-2} dt = du$$

So  $\widehat{B}_u$  is a standard Brownian motion.

$$\text{Let } \widehat{F}(x+iy) := \left(1 + \left(\frac{x}{y}\right)^2\right)^{\frac{\theta}{2}} y^{-\lambda}$$

$$M_u := \psi(u)^{\alpha} \widehat{F}(z(u)).$$

$$\text{By It\^o: } dM_u = -2M \frac{bx}{x^2+y^2} ds = \sqrt{2\kappa} M \frac{bx}{\sqrt{x^2+y^2}} d\widehat{B}$$

So  $M_u$  is a local martingale.

$$\langle M, M \rangle_u = 2\kappa \int_0^u M_s^2 \frac{b^2 x_s^2}{x_s^2 + y_s^2} ds \leq c(T) \int_0^u (M_s^2 + 1) ds \text{ as } \log z_s \leq T.$$

By a Lemma we had before,  $M$  is a martingale.

$$\text{So } \psi(u) \widehat{F}(z) = E(\psi(0)^{\alpha} \widehat{F}(z(0))) = \widehat{F}(z)$$

$$\widehat{F}(z) = E(\psi(u)^{\alpha} \widehat{F}(z))$$

**Lemma (The Key Estimate)** Let  $\theta \in [0, 1 + \frac{\kappa}{2}]$ ,

$$\alpha = 2\theta + \frac{\kappa\theta - \kappa}{2}, \quad \lambda = \theta + \kappa(1-2\theta)/2.$$

Then  $\exists c(\kappa, \theta): \forall t \in [0, 1], y > 0, s \in [0, 1], x \in \mathbb{R}$ .

$$P(|\widehat{F}_t^{\uparrow}(x+iy)| > \frac{\delta}{y}) \leq c(\kappa, \theta) \left(1 + \frac{x^2}{y^2}\right)^{\frac{\lambda}{2}} \left(\frac{y}{\delta}\right)^{\lambda} A(\delta, \alpha - \lambda)$$

where

$$A(\delta, \mu) := \begin{cases} \delta^{-\mu}, & \mu > 0 \\ 1 + \log \delta, & \mu = 0 \\ 1, & \mu < 0 \end{cases}$$

Proof. Assume  $\delta > y$ , otherwise RHS is  $\geq c(\kappa, \theta)$ .

and we can take  $c(\kappa, \theta) \geq 1$ .

$\widehat{F}'_t(z)$  has the same distribution as  $g'_t(z)$ .

Denote:  $u_t := \log(\text{Im } g_t(x+iy))$

Observe:  $\left| \partial_x \log |g'_t(z)| \right| = \frac{\operatorname{Re}((g_t(z) - f(t))^2)}{|g_t(z) - f(t)|^2} \leq 1$

So for any  $u$ ,  $\left| \frac{g'_{-t}(z)}{g'_{T_j}(z)} \right| \leq \exp(|u - u_1|)$ .

So  $P\left(|g'_{-t}(z)| \geq \frac{\delta}{y}\right) \lesssim \sum_{j=\lceil \log y \rceil}^0 P\left(|g'_{T_j}(z)| \geq \frac{\delta}{y}\right)$

All the constants in " $\geq$ " and " $\lesssim$ " are independent on  $y$  and  $t$  because  $|u_1| = |\log |\operatorname{Im} g_{-t}(z)|| \leq c_1$  for some universal  $c_1$ , since  $y, t \leq 1$  (and thus  $\operatorname{Re} g_{-t} \leq 2$ )

By Schwarz lemma,  $y |g'(z)| \leq \operatorname{Im} g(z)$  ( $y: \mathbb{H} \rightarrow \mathbb{H}$ )

So  $P\left(|g'_{-t}(z)| \geq \frac{\delta}{y}\right) \lesssim \sum_{j=\lceil \log y \rceil} P\left(|g'_{T_j}(z)| \geq \frac{\delta}{y}\right)$   
 enough to start at  $\lceil \log y \rceil$ .

Now, we can use scale invariance to get that if  $j \in \lceil \log y, 0 \rceil$ , then

$$E\left(y^a e^{-ja} |g'_{T_j}(z)(z)|^a\right) \leq E(e^{-jz})$$

So  $P\left(|g'_{T_j}(z)(z)| \geq \frac{\delta}{y}\right) = P\left(|g'_{T_j}(z)|^a y^a \delta^{-a} \geq 1\right) \stackrel{\text{Chebyshev}}{\leq}$

$$E\left(|g'_{T_j}(z)|^a y^a \delta^{-a}\right) \leq \delta^a e^{ja} E(e^{-jz})$$

So  $P\left(|g'_{-t}(z)| \geq \frac{\delta}{y}\right) \leq \left(1 + \frac{x^2}{y^2}\right)^{\delta} \delta^{-a} y^{\lambda} \sum_{j=\lceil \log y \rceil}^0 e^{j(\alpha - \lambda)}$

Which gives the lemma.  $\square$