

Brownian Motion and Analytic Functions

by

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Abstract

This paper is mostly expository and is concerned with the connection between two dimensional Brownian motion and analytic functions provided by P. Lévy's result that, if Z_t , $0 \leq t < \infty$, is two dimensional Brownian motion, and if f is analytic and not constant, then $f(Z_t)$, $0 \leq t < \infty$, is also two dimensional Brownian motion, perhaps moving at a different speed. This can be used to study Brownian motion via analytic functions and, conversely, to treat analytic functions probabilistically. Recently several open problems in analytic function theory have been solved in this manner. We will present some of Doob's earlier work on the range and boundary values of analytic functions, the probabilistic theory of H^p spaces due to Burkholder, Gundy, and Silverstein, the author's results on conjugate function inequalities, and sketch probabilistic proofs of Picard's big and little theorems, Koebe's theorem, and others. There are some new results.

1. Introduction. A fundamental connection between two dimensional Brownian motion $Z_t = X_t + iY_t$, $t \geq 0$, moving in the complex plane, and analytic function theory, is Paul Lévy's theorem that, if $f(z)$ is analytic and not constant, the process $f(Z_t)$, $t \geq 0$, is again Brownian motion, although perhaps moving at a different speed than Z_t . This conformal invariance can be used to study Brownian motion via analytic functions, by making a judicious choice of $f(z)$. For example ([16]) let $Z_0 \equiv 0$ and let $a \neq b$ be complex numbers. Then $(a-b)e^{Z_t} + b$, $t \geq 0$, is Brownian motion started at $(a-b)e^0 + b = a$. Clearly it never hits b , since e^z never vanishes. This proves the well known result that the probability Brownian motion ever hits a fixed point other than its starting point is 0. (Here Brownian motion means two dimensional Brownian motion moving in the complex plane, the mathematical description by Norbert Wiener of a physical process observed by, among others, Brown.)

Levy's result can also be used to study analytic functions probabilistically, the principal subject of this paper. In this context it is usually applied to a collection of functions. For example, it is proved in the next section that Brownian motion hits each closed set of positive capacity with probability one. Thus, the range of each nonconstant entire function cannot omit a closed set of positive capacity, since the Brownian motion $f(Z_t)$ moves entirely in $f(\mathbb{C})$.

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Except for new material in the last section, this paper is an expository presentation of some of the applications of the conformal invariance of Brownian motion to analytic function theory, including the recent solutions of several open problems. Often probabilistic expressions can be written in classical terms, and the theorems of probability can be brought to bear directly on non-probabilistic problems. Other expressions arise which do not seem amenable to such a translation, and in this way intrinsically new tools arise, which, aside from their use in the study of classical function theory, are sometimes interesting enough to deserve attention in their own right, and are thus a source of new theorems and problems. Several of the latter are given at the end of this paper.

Brownian motion is a very intuitive thing to work with, because at the back of our mind is the corresponding physical process, or rather an idealized version of it, which is only very briefly described at the beginning of the next section. A much more detailed and very enjoyable physical and historical account of Brownian motion can be found at the beginning of Edward Nelson's book, [33].

Although there is necessarily some overlap between the subject of this paper and the better known probabilistic potential theory, no attempt will be made to treat the latter systematically. See the book [21] of E. F. Dynkin and A. A. Yuskevitch for a very readable elementary introduction to the use of Brownian motion in studying harmonic and subharmonic functions, and see [18] for more advanced treatment.

For simplicity, the treatment here is usually restricted to entire functions and functions defined on the unit disc. Extensions of the results here to more general situations will often be immediate. Proofs are usually just sketched, except in the next section, where the

foundations of the subject, up to and including some basic theorems of Kakutani and a version of Lévy's theorem, are rigorously presented.

In Section three the author's proof of Picard's little theorem is sketched. This theorem can be proved using Brownian motion on Riemann surfaces and the modular function, together with ideas of Kakutani (see Section 8). Here Riemann surfaces are not used and the work of the modular function is done by a law of large numbers, which is used to derive a result about Brownian motion paths originally proved (via the modular function) by Ito and McKean. For another, shorter, exposition of some aspects of this proof see P. J. Kahane's paper [28]. We remark that Picard used the modular function in his proof and that the first elementary (that is, modular function free) proof were based on Bloch's theorem. Now there are a number of proofs.

In Section four some of Doob's earlier work on the range and boundary values of analytic functions is sketched. One sample of this has already been given. The particularly close connection between Stoltz regions and Brownian paths is discussed here. See Burkholder, [6], for related expository material.

The next topic is the theorems and techniques associated with the probabilistic treatment of H^p spaces of analytic functions, including the famous solution by D. L. Burkholder, R. F. Gundy, and M. L. Silverstein of a long standing problem in this area in [12], and some of Burkholder's later work. The original proof, which used martingales, is translated to enable use of Lévy's theorem. This is the approach taken by Burkholder more recently. Karl Peterson has written a book, [34], mostly devoted to

presenting and explaining [12]. For another exposition of [12] as well as a thorough treatment of modern H^p theory, much of which was inspired by [12], see Charles Fefferman's paper [22]. Fefferman and E. Stein were the first to give a non-probabilistic proof of the Burkholder, Gundy, and Silverstein result, in [23].

In Section 6 the author's method of studying conjugate function and Hilbert transform inequalities will be sketched. This provides a uniform approach to a number of these inequalities and often gives the extremal functions and best possible constants for them, some of which were found for the first time in this manner. There are expository accounts of this method, applied to different problems than the one considered here, in J. P. Kahane's paper [28] and D. L. Burkholder's paper [6]. Recently Albert Baernstein II has given non-probabilistic proofs of most of these results, as well as some very nice extensions, in [2].

The next to last section contains new proofs of Koebe's theorem (without the constant $1/4$) and Picard's big theorem, and in the final sections some problems are posed and Brownian motion on Riemann surfaces, including an early paper of Kakutani on the type problem for Riemann surfaces, briefly discussed.

2. Fundamentals. A very small particle suspended in a liquid can be seen to move rapidly about, due to the bombardment of the particle by the molecules of the liquid. If the position of the particle is projected onto a plane, a two dimensional motion results. This is what would be observed if the particle was watched through a microscope and it was impossible to discern the up and down component of its position.

In the 1920's Norbert Wiener gave a mathematical description of an idealized version of this motion. We will distinguish between standard Brownian motion, supposed to describe the movement of a particle suspended in a liquid of a certain unchanging temperature, and Brownian motion, in which the temperature of the liquid, which influences the rapidity of the motion of the particle, is allowed to vary.

Wiener constructed a family of random variables $Z_t = X_t + iY_t$, $t \geq 0$, on a probability space Ω , Z_t representing the position of the particle at time t . For each $\epsilon > 0$, let $\tau_0(\epsilon) = \tau_0 = 0$, and if $i \geq 1$ define $\tau_i(\epsilon) = \tau_i = \inf\{t > \tau_{i-1} : |Z_t - Z_{\tau_{i-1}}| = \epsilon\}$, and let $\Delta_i(\epsilon) = \Delta_i = Z_{\tau_i} - Z_{\tau_{i-1}}$. The process Z_t satisfies the following postulates.

- (A) For almost every $\omega \in \Omega$, the path $Z_t(\omega)$, $0 \leq t < \infty$, is continuous and unbounded.
- (B) For each $\epsilon > 0$, $\Delta_1, \Delta_2, \dots$ are independent and each is uniformly distributed with respect to linear Lebesgue measure on $\{|z| = \epsilon\}$.
- (C) For each $\epsilon > 0$, $\tau_i - \tau_{i-1}$, $i \geq 1$, are independent and identically distributed. Furthermore $E(\tau_i(1) - \tau_{i-1}(1)) = 1/2$ (a normalization).

Standard Brownian motion is defined to be a process satisfying these three postulates, while Brownian motion is defined to be a process satisfying the first two. It is not difficult to show that this definition of standard Brownian motion is equivalent to the usual one. What we call Brownian motion might be given other names elsewhere. Wiener's construction of standard Brownian motion may be found in [4], and there is a different construction in [32].

Although these definitions are new, it has been recognized for a long time that the property of hitting circles with a uniform distribution is very useful when using Brownian motion to study harmonic and analytic functions. Of course postulate (B) means that, for each integer n , the

measure on \mathbb{C}^n induced by the map $(\Delta_1, \dots, \Delta_n)$ from Ω is product measure μ^n , where μ puts uniform measure on the circle of radius ϵ about 0.

In the rest of this paper Z_t will be standard Brownian motion, and P_z and E_z will denote probability and expectation associated with Z_t started at z , that is satisfying $Z_0 = z$. Since Z_t will usually start at 0, P_0 and E_0 will be written P and E to avoid subscripts. Linear Lebesgue measure will be denoted by λ , and $a \wedge b = \min(a, b)$.

Recall that a harmonic function defined on a region R is a continuous function which satisfies the averaging property. That is if $\{|z - z_0| \leq \epsilon\} \subset R$ then

$$(2.1) \quad u(z_0) = \int_0^{2\pi} u(z_0 + \epsilon e^{i\theta}) d\theta / 2\pi.$$

This connects nicely with postulate (B). The following fundamental theorem is due to Kakutani, in [29]. The proof is essentially Doob's ([18]). Although it is stated only for standard Brownian motion, since it is only a theorem about the paths of Z_t , it also holds for (non-standard) Brownian motion. It will be seen that only postulates (A) and (B) are used in the proof. This also is true of Theorem 2.2 and 2.3.

Theorem 2.1. Let u be harmonic on a bounded region R and continuous on \bar{R} (closure of R). Let $\tau_R = \inf\{t \geq 0: Z_t \in \partial R \text{ (boundary of } R)\}$. Then $E_z u(Z_{\tau_R}) = u(z)$, $z \in R$.

Proof. Let $\epsilon > 0$ be fixed for a while, and let $N = \min\{k: |Z_{\tau_k} - \partial R| \leq \epsilon\}$, where $\tau_k = \tau_k(\epsilon)$ is defined as before.

If $|z - \partial R| \leq \epsilon$, $N = 0$, so assume $|z - \partial R| > \epsilon$ which gives $P(N \geq 1) = 1$.

Then

$$E_z u(Z_{\tau_1}) = \int_0^{2\pi} u(z + \epsilon e^{i\theta}) d\theta / 2\pi = u(z)$$

by (2.1) and (B).

Now $Z_{\tau_2} - Z_{\tau_1}$ is uniformly distributed on $\{|z| = \epsilon\}$, and independent of $Z_{\tau_1} - Z_{\tau_0}$, so it is independent of $\{N > 1\}$. Thus, integrating first with respect to the second coordinate and then the first in $\mathbb{C} \times \mathbb{C}$, (2.1) and (B) again give:

$$\begin{aligned} & E_z (u(Z_{\tau_2}) - u(Z_{\tau_1})) I(N > 1) \\ &= E_z \left[\frac{1}{2\pi} \int_0^{2\pi} u(Z_{\tau_1} + \epsilon e^{i\theta}) d\theta - u(Z_{\tau_1}) \right] I(N > 1) \\ &= E_z [0] \cdot I(N > 1) = 0, \text{ so} \end{aligned}$$

$$E_z u(Z_{\tau_2 \wedge N}) = E_z u(Z_{\tau_1}) + E_z (u(Z_{\tau_2}) - u(Z_{\tau_1})) I(N > 1) = u(z).$$

Continuing in this manner we get $E_z u(Z_{\tau_N \wedge k}) = u(z)$ for each k , and the bounded convergence theorem gives $E_z u(Z_{\tau_N}) = u(z)$. Now (A) gives $Z_{\tau_N} \rightarrow Z_{\tau_R}$ almost surely as $\epsilon \rightarrow 0$, so $u(Z_{\tau_N}) \rightarrow u(Z_{\tau_R})$ a.s., and another application of the bounded convergence theorem completes the proof.

This theorem implies immediately that the distribution of Z_{τ_R} under P_z , and harmonic measure on ∂R with respect to the region R and the point z , are the same. Readers not familiar with the concept of harmonic measure may use this fact to give a probabilistic definition. Sometimes probabilistic arguments may be used to find or estimate harmonic measure when other methods don't work. Conversely, Brownian hitting probabilities can be found by using the standard methods to find harmonic measure, as

in what follows.

Let $0 < a < 1 < A < \infty$ and let R be the annulus $\{a < |z| < A\}$, and apply Theorem 2.1 to the function $u(z) = \ln|z|$. Let $p(a, A) = p = P_1(Z_{\tau_R} = a)$ so that $1-p = P_1(Z_{\tau_R} = A)$. Theorem 2.1 gives

$$(2.2) \quad 0 = \ln(1) = E_1 \ln|Z_{\tau_R}| = p \ln(a) + (1-p) \ln(A).$$

If A is held fixed and $a \rightarrow 0$, (2.2) gives $p \rightarrow 0$, which proves, since A was arbitrary and Z_t has continuous paths a.s., that the probability Z_t ever hits $\{0\}$ is 0. This proof goes through similarly if Z_t starts at any point except 0, and uses only the existence of a harmonic function on $\mathbb{C} - \{0\}$ which goes to ∞ at 0 and to $-\infty$ at ∞ . An analogous function exists for every compact set of capacity 0, so the probability Brownian motion started at a point outside such a set ever hits it is 0.

If we hold a fixed and let $A \rightarrow \infty$ in (2.2), then $p \rightarrow 1$. Thus Brownian motion started at 1 hits $\{|z| = a\}$ with probability 1, for each $a > 0$, and it is easy to argue (using the Strong Markov Property, which will soon be discussed) that $\{|z| = a\}$ is almost surely visited at arbitrarily large times t , or, to put it another way, the probability that Z_t , $t \geq n$, visits $\{|z| = a\}$ is one for each integer n . Any compact set of positive capacity may be handled similarly, giving the following theorem of Kakutani, ([29]), which can be used to give a definition of capacity 0.

Theorem 2.2. Let K be a compact subset of \mathbb{C} .

If $\text{Cap}(K) = 0$, $P_z(Z_t \in K \text{ for some } t \geq 0) = 0$, $z \notin K$.

If $\text{Cap}(K) > 0$, $P_z(Z_t \in K \text{ for arbitrarily large } t) = 1$, for all z in \mathbb{C} .

A nonnegative random variable T will be called a Markov time for Z_t if $Z_{T+t} - Z_T$, $t \geq 0$ is a standard Brownian motion independent of the σ -field $\sigma(Z_{t \wedge T}, 0 \leq t < \infty)$ (the past up to time T). A nonnegative

random variable T is called a stopping time for Z_t if, for each $\lambda > 0$, $\{\tau < \lambda\}$ is in this σ -field. Thus, if τ is a stopping time, whether stopping has occurred by time λ can be determined by observing Z_t up to time λ . For instance, the first exit time from a region is a stopping time. It can be proved that all stopping times for standard Brownian motion are Markov times (see [4]). Another way to say this is that standard Brownian motion has the Strong Markov Property.

Now a version of Lévy's theorem will be proved. In its simplest form, which is sufficient for many applications, this theorem says the following.

Theorem 2.3. If $f(z)$ is a nonconstant entire function, and Z_t starts at z_0 , then $f(Z_t)$, $0 \leq t < \infty$, is Brownian motion starting at $f(z_0)$.

Proof. Clearly, $f(Z_t)$ has continuous paths a.s., since Z_t does, and they are almost surely unbounded since, by Theorem 2.2, Z_t visits $\{|f(z)| \geq \lambda\}$ with probability one. This takes care of (A).

Let $\gamma_1 = \inf\{t > 0: |f(Z_t) - f(z_0)| = \epsilon\}$. To show that $f(Z_{\gamma_1})$ is uniformly distributed on $\{|z - f(z_0)| = \epsilon\}$ it is sufficient to show that, for all functions u harmonic on $S = \{z: |z - f(z_0)| < \epsilon\}$, and continuous on \bar{S} , $E_{z_0} u(f(Z_{\gamma_1})) = u(f(z_0))$. This is an immediate consequence of Theorem 2.1, since, if R is the component of $\{z: |f(z) - f(z_0)| < \epsilon\}$ containing z_0 , $u(f(z))$ is harmonic on R and continuous on \bar{R} , while $\gamma_1 = \inf\{t > 0: Z_t \in \partial R\}$.

Now if, $\gamma_2 = \inf\{t > \gamma_1: |f(Z_t) - f(Z_{\gamma_1})| = \epsilon\}$, $f(Z_{\gamma_2}) - f(Z_{\gamma_1})$ is uniformly distributed on $\{|z| = \epsilon\}$ and independent of $f(Z_{\gamma_1}) - f(z_0) = f(Z_{\gamma_1}) - f(z_0)$. This follows immediately from the Strong Markov Property, the fact that γ_1 is a stopping time for Z_t , and the result of the previous paragraph. An iteration of this argument gives property (B) for $f(Z_t)$.

Several things may now be evident. We have only made use of property (B) in the limit as $\epsilon \rightarrow 0$. In fact, (B) for only very small ϵ can be shown to imply (B) for all ϵ . This has the corollary that whether a process is a Brownian motion is largely a local property. Thus, an heuristic proof of Lévy's theorem could be given as follows.

Locally, analytic functions are almost like $az + b$. The functions $az + b$ clearly take Brownian motion to Brownian motion, Q.E.D. This argument can be made rigorous, too. See McKean's book [32], page 109.

Now a more complicated version of Lévy's theorem will be stated but not proved. See [32] for a proof. The following notation will be used throughout the paper. If $f(z)$ is analytic and not constant in the unit disc D , and $Z_0 \equiv 0$, define

$$(2.3) \quad \rho_f(s) = \rho(s) = \int_0^s |f'(Z_t)|^2 dt, \quad 0 \leq s < \tau_D.$$

Since Z_t misses the (countable) zeros of f' , ρ is almost surely strictly increasing. Now $f(Z_t)$ is Brownian motion, but perhaps moves locally too fast or slow to be standard Brownian motion. We speed it up or slow it down by changing the time scale. Let

$$(2.4) \quad W_t = f(Z_{\rho^{-1}(t)}), \quad 0 \leq t < \tau_D.$$

Then W_t is locally standard Brownian motion, but

it is only defined up to $t = \rho(\tau_D)$, which may or may not be infinite.

The process W_t could be talked about as standard Brownian motion up to time $\rho(\tau_D)$, but it is more convenient to define W_t for $t \geq \rho(\tau_D)$ so that the whole process W_t , $0 \leq t < \infty$, is standard Brownian motion.

This is done by defining, on $\{\rho(\tau_D) < \infty\}$,

$$(2.5) \quad W_{\rho(\tau_D)} = \lim_{t \rightarrow \tau_D} W_{\rho(t)} \quad \text{and}$$

$$W_{\rho(\tau_D)+t} = W_{\rho(\tau_D)} + (Z_{\tau_D+t} - Z_{\tau_D}), \quad t > 0.$$

The limit can be shown to exist almost surely on $\{\rho(\tau_D) < \infty\}$. One form of Lévy's theorem states.

Theorem 2.4. The process W_t , $0 \leq t < \infty$, is standard Brownian motion.

Because of the central role of the time $\rho_f(\tau_D)$ in what follows we designate

$$\rho_f(\tau_D) = \nu(f)$$

which will be further shortened to ν if the function f is clear from the context. The distribution of ν is a measure of the size of $f(D)$. For example, if f is univalent and maps D onto R and 0 to 0 , then $P(\nu > a)$ is the probability that standard Brownian motion started at 0 takes more time than a to exit from R .

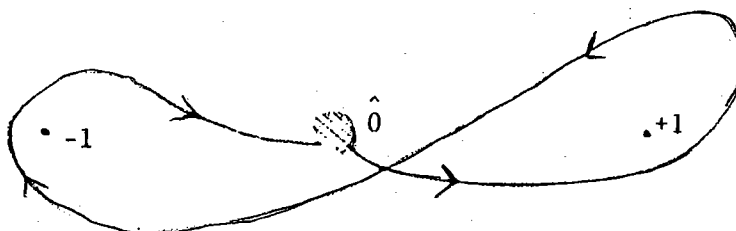
3. Picard's little theorem. This theorem, which states that if f is a nonconstant entire function then the range of f contains every complex number, except perhaps one, is of course equivalent to the statement that if a and b are distinct complex numbers then either a or b is in $f(\mathbb{C})$. It involves no loss of generality to assume $a = 1$ and $b = -1$, and we make the further simplifying assumption, which can be easily circumvented, that $f(0) = 0$. What will be shown, then, is that, if $f(z)$ is a nonconstant entire function satisfying $f(0) = 0$, then one of the points ± 1 is in $f(\mathbb{C})$. Assume, to the contrary, that

$f(C)$ contains neither of these points. Identify the points of $\{z: |z| \leq .1\}$, call this set $\hat{0}$, and let $\hat{\mathbb{C}}$ be $\mathbb{C} - \{+1, -1\}$ with the points of $\hat{0}$ identified. Let \mathcal{O} be the component of $f^{-1}(\hat{0})$ containing 0, and let \mathcal{C} be \mathbb{C} with the points of \mathcal{O} identified. Then f gives a continuous map from \mathcal{C} to $\hat{\mathbb{C}}$, in the usual manner, and a closed curve in \mathcal{C} is mapped to a closed curve in $\hat{\mathbb{C}}$. Clearly any closed curve in \mathcal{C} can be continuously shrunk, while being kept a closed curve, down to a single point, while this is not true of all closed curves in $\hat{\mathbb{C}}$, only those which are not tangled around ± 1 (i.e., those homotopic to 0). Thus the image under f of any closed curve K in \mathcal{C} must be a closed curve in $\hat{\mathbb{C}}$ which is homotopic to 0, since, as K shrinks to a point, so must $f(K)$. A contradiction will be gotten by exhibiting a closed curve in \mathcal{C} with an image not homotopic to 0 in $\hat{\mathbb{C}}$. This curve will be (the projection of) a Brownian motion path.

First we need the following Lemma, equivalent to a theorem of Ito and McKean (see [32]), who used the modular function in their proof. A different proof is indicated.

Theorem 3.1. There is a time τ , $P(\tau < \infty) = 1$, such that $s \geq \tau$ and $Z_s \in \hat{0}$ implies Z_t , $0 \leq t \leq s$, is not homotopic to 0 in $\hat{\mathbb{C}}$.

The tangling of a curve in $\hat{\mathbb{C}}$ can be represented by a word written using the four "letters" a , a^{-1} , b , and b^{-1} , where a and a^{-1} stand respectively for clockwise and counterclockwise loops around 1 and b and b^{-1} serve similarly for -1 . The curve in the picture would have $a^{-1}b$ as its word.



If this curve is a Brownian motion path, it will get less tangled in the future only if it loops counterclockwise around -1 , i.e. unwinds, before it does any of the other three possibilities, which would tangle it more. Each of the four possibilities is about equally likely, by symmetry, so the Brownian motion is about three times as likely to become more tangled up as less. By considering times $\gamma_1 < \gamma_2 < \gamma_3 \dots \rightarrow \infty$ such that $Z_{\gamma_i} \in \hat{0}$, and such that $Z_t, 0 \leq t \leq \gamma_{i+1}$, is about three times as likely to be more tangled up in $\hat{0}$ than $Z_t, 0 \leq t \leq \gamma_i$, and by using a law of large numbers related to the one which says that, if a coin with probability $3/4$ heads and $1/4$ tails is tossed repeatedly, eventually heads become and stay more numerous than tails, Theorem 3.1 can be proved.

Note that although Theorem 3.1 is stated for standard Brownian motion, since it involves only path properties it is also true for Brownian motion.

Now the proof of Picard's Theorem can be completed. Let $\tau(f)$ be the τ guaranteed by Theorem 3.1 for the Brownian motion $f(Z_t)$. Since $Z_t \in \hat{0}$ for arbitrarily large times t by Theorem 2.2, there is a time $\eta > \tau(f)$, $P(\eta < \infty) = 1$, such that $Z_\eta \in \hat{0}$, which implies $f(Z_\eta) \in \hat{0}$. Any of the paths $Z_t, 0 \leq t \leq \eta$, gives the contradiction. Details of the above can be found in [16].

4. Stoltz angles. The idea to use probability to help prove theorems of the type proved in this section is Doob's, and he worked in a **very general setting**. His proofs, which were based in part on classical theorems in potential theory, especially those involving the fine topology, have been altered here to make use of recent techniques of Burkholder, Gundy, and Silverstein for dealing probabilistically with Stoltz domains.

First, two more or less immediate theorems concerning Brownian motion and analytic functions will be proved, and then the older classical analogues will be stated, together with a sketch of how to get from the former to the latter. If f is a nonconstant entire function and H is a closed set then, using the notation introduced at the end of Section 2, $\{W_\nu \in H\} = \{\lim_{t \rightarrow \tau_D} f(Z_t) \in H\}$ using (2.4) and (2.5). Theorem 2.2 now immediately gives the following.

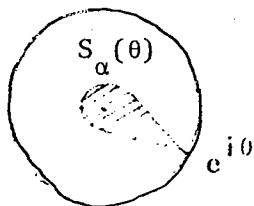
Theorem 4.1 If f is a nonconstant entire function, and K is a compact set of capacity 0, then

$$P(\lim_{t \rightarrow \tau_D} f(Z_t) \in K) = 0.$$

The other half of Theorem 2.2 implies that, for each number a , almost every path W_t , $a \leq t < \infty$, is dense in the plane, since the probability it visits each disc of rational center and radius is one. Thus, the probability W_t , $a \leq t < \infty$, is dense in \mathbb{C} for all a is one, which implies that on $\{\nu = \infty\}$, almost all the paths $f(Z_t)$, $(\tau_D - \epsilon) \wedge \tau_D \leq t < \tau_D$, are dense in \mathbb{C} for all $\epsilon > 0$, since W_t , $0 \leq t < \nu$, and $f(Z_t)$, $0 \leq t < \tau_D$, traverse the same paths at different speed. On $\{\nu < \infty\}$, $\lim_{t \rightarrow \tau_D} f(Z_t)$ exists and equals W_ν almost surely. This gives the following dichotomy.

Theorem 4.2. With probability 1, either $\lim_{t \rightarrow \tau_D} f(Z_t)$ exists or $f(Z_t)$, $\tau_D - (\epsilon \wedge \tau_D) < t < \tau_D$, is dense in the plane for each $\epsilon > 0$.

The classical analogues of these theorems involve Stoltz domains, which will now be defined. For each $e^{i\theta}$, and each α between 0 and 1, define the Stoltz domain $S_\alpha(\theta)$ to be the interior of the smallest convex set containing the disc $\{|z| \leq \alpha\}$ and the point $e^{i\theta}$ (see the picture).



If A is a subset of ∂D , define $S_\alpha(A) = \bigcup_{e^{i\theta} \in A} S_\alpha(\theta)$. These Stoltz domains often have sawtooth like boundaries. The following theorem is implicit in [20], although the proof is based on a result of Naim about fine limits. A probabilistic proof is implicit in the last half of Burkholder's paper, [6].

Theorem 4.3. For each fixed α , $0 < \alpha < 1$, and each Borel set A , $P(Z_t \in S_\alpha(A)), \tau_D - (\epsilon \wedge \tau_D) < t < \tau_D$, and $Z_{\tau_D} \in A \rightarrow P(Z_{\tau_D} \in A)$ as $\epsilon \rightarrow 0$.

This theorem is of course vacuous if $\lambda(A) = 2\pi P(Z_{\tau_D} \in A) = 0$, but, if $P(A) > 0$, it says that almost all Brownian paths which hit points in A get in and stay in $S_\alpha(A)$ before they hit. It will not be proved, but note that it is immediate if A is open.

A function g , analytic or not, defined in D , is said to have a nontangential limit at $e^{i\theta}$ if

$$\lim_{\substack{z \rightarrow e^{i\theta} \\ z \in S_\alpha(\theta)}} g(z) \text{ exists for all } \alpha, 0 < \alpha < 1.$$

Using Theorem 4.3 it is easy to show that, if g is any function with a nontangential limit $\hat{g}(e^{i\theta})$ on a Borel set $B \subset \partial D$,

$$(4.1) \quad P(\lim_{t \rightarrow \tau_D} g(Z_t) = \hat{g}(Z_{\tau_D}), Z_{\tau_D} \in B) = P(Z_{\tau_D} \in B).$$

This equality and Theorem 4.1 now give the following theorem of Privilov.

Theorem 4.4. If f is a nonconstant entire function, and K is a compact set of capacity 0, then the Lebesgue measure of the set of ∂D where f has nontangential limit in K is 0.

Now let θ be fixed. The concept of Brownian motion Z_t conditioned so that $Z_{\tau_D} = e^{i\theta}$ is intuitive, but difficult to make rigorous since $P(Z_{\tau_D} = e^{i\theta}) = 0$. Nonetheless this was done by Doob in [19] and the resulting process will be denoted Z_t^θ , $0 \leq t \leq \tau_D$. Here, Z_0^θ will always be 0. As you would expect, if E a Borel set in D such that $e^{i\theta} \notin \bar{E}$, and $\Gamma_\epsilon = \{e^{i\varphi} : \theta - \epsilon \leq \varphi \leq \theta + \epsilon\}$,

$$(4.2) \quad P(Z_t^\theta \in E \text{ for some } t, 0 \leq t \leq \tau_D) \\ = \lim_{\epsilon \rightarrow 0} \frac{P(Z_t \in E \text{ for some } t, 0 \leq t \leq \tau_D, \text{ and } Z_{\tau_D} \in \Gamma_\epsilon)}{P(Z_{\tau_D} \in \Gamma_\epsilon)},$$

and the probabilities of other events for Z_t^θ are similarly computed.

(If $e^{i\theta} \in \bar{E}$, the limit may not exist.) Conditional Brownian motion is nice enough to be used in the following way. If E is a Borel subset of D ,

$$(4.3) \quad P(Z_t \in E \text{ for some } t, 0 \leq t < \tau_D) \\ = \int_0^{2\pi} P(Z_t^\theta \in E \text{ for some } t, 0 \leq t \leq \tau_D) d\theta/2\pi.$$

(If $\bar{E} \cap \partial D = \emptyset$, (4.3) can be derived directly from (4.2) without much difficulty)

A fundamental connection between conditional Brownian motion and Stoltz angles is the following theorem, essentially proved in [12].

Theorem 4.5. For each σ , $0 < \sigma < 1$, there is a constant $a(\sigma) > 0$ such that, for all $z \in S_\sigma(\theta)$,

$$P(Z_t^\theta, 0 \leq t < \tau_D, \text{ contains a closed loop around } z) \geq a(\sigma).$$

Now wrapping around a point in a connected set F means intersecting the set, so if $F \cap S_\sigma(\theta)$ contains points $z_n \rightarrow e^{i\theta}$, and if A_n is the event that Z_t^θ , $0 \leq t < \tau_D$, contains a closed loop around z_n , then

$P(\overline{\lim} A_n) \geq \overline{\lim} P(A_n) \geq a_\sigma$. From this it can be concluded, since

Z_t^θ , $0 \leq t \leq \tau_D$, is a curve which first hits ∂D at τ_D , that

$$(4.4) \quad P(Z_t^\theta, 0 \leq t < \tau_D, \text{ hits } F \text{ at times arbitrarily close to } \tau_D) \geq a(\sigma),$$

and a 0-1 law ([3], page 30), applied to the reversed process, gives

$$(4.5) \quad P(Z_t^\theta, 0 \leq t < \tau_D, \text{ hits } F \text{ at times arbitrarily close to } \tau_D) = 1.$$

Now let u be a harmonic function, and define $u_\sigma^+(e^{i\theta}) = \overline{\lim}_{\substack{z \rightarrow e^{i\theta} \\ z \in S_\sigma(\theta)}} u(z)$.

Applying (4.5) to the connected sets $\{u > \lambda\}$ gives

$$P(\overline{\lim}_{t \rightarrow \tau_D} u(Z_t^\theta) \geq u_\sigma^+(e^{i\theta})) = 1.$$

and integrating with respect to $d\theta$ gives

$$(4.6) \quad P(\overline{\lim}_{t \rightarrow \tau_D} u(Z_t) \geq u_\sigma^+(Z_{\tau_D})) = 1.$$

There is a corresponding result for lim.

Now let $T_\sigma = \{e^{i\theta} : \lim_{\substack{z \rightarrow e^{i\theta} \\ z \in S_\sigma(\theta)}} f(z) \text{ does not exist}\}$. Either the

limit of the real or imaginary part of f fails to exist as $z \rightarrow e^{i\theta}$ in $S_\sigma(\theta)$, $e^{i\theta} \in T_\sigma$, so that (4.6) and its analogue for lim give

$$P(\lim_{t \rightarrow \tau_D} f(Z_t) \text{ exists}, Z_{\tau_D} \in T_\sigma) = 0,$$

which implies

$$(4.7) \quad P(v = \infty, Z_{\tau_D} \in T_\sigma) = P(Z_{\tau_D} \in T_\sigma) = \lambda(T_\sigma)/2\pi,$$

while (4.1) gives

$$(4.8) \quad P(v < \infty, Z_{\tau_D} \notin T_\sigma) = P(Z_{\tau_D} \notin T_\sigma).$$

Equation (4.7), Theorem 4.3, and the second sentence before the statement of Theorem 4.2, give that, if A is any Borel subset of T_σ , then $f(S_\sigma(A))$ is dense in \mathbb{C} , which implies that $f(S_\sigma(\theta))$ is dense in \mathbb{C} for almost every $\theta \in T_\sigma$. Furthermore, (4.7) and (4.8) imply $\ell(T_\sigma)$ is the same ($2\pi P(v = \infty)$) for all σ . Thus the following theorem of Plessner holds.

Theorem 4.5. Except for a set of θ of Lebesgue measure 0, either f has a nontangential limit at $e^{i\theta}$ or $f(S_\alpha(\theta))$ is dense in \mathbb{C} for each α , $0 < \alpha < 1$.

5. Hardy spaces. The probabilistic treatment of H^p spaces, by D. L. Burkholder, R. F. Gundy, and M. L. Silverstein, was made possible by Burkholder and Gundy's proof, a few years earlier, of a theorem about standard one dimensional Brownian motion. If $Z_t = X_t + iY_t$ is standard two dimensional Brownian motion then X_t and Y_t are independent standard one dimensional Brownian motions. This connection is usually used to define two dimensional Brownian motion, in fact. The definition of Markov time for standard one dimensional Brownian motion parallels the definition of Markov time given in Section 2, and if τ is a Markov time for Z_t then it is for X_t and Y_t also. When $P(\tau = \infty) > 0$, we say that τ is a Markov time if $\tau \wedge n$ is a Markov time for each integer n . It follows from the construction of Section 2 that ν is a Markov time for W_t and thus for $\text{Re } W_t$ and $\text{Im } W_t$. If A_t , $0 < t < \infty$, is any stochastic process, define the new process A_t^* , $0 \leq t < \infty$, by $A_t^* = \sup_{0 \leq s \leq t} |A_s|$. In [11], Burkholder and Gundy prove that, if Γ_t , $0 \leq t < \infty$, is standard one dimensional Brownian motion satisfying $\Gamma_0 = 0$, then, for each $p > 0$, there are positive constants c_p and C_p such that

$$(5.1) \quad c_p E\tau^{p/2} \leq E\tau^{*p} \leq C_p E\tau^{p/2}$$

for all Markov times τ .

Let \mathfrak{F} be the class of all functions $f = u + iv$, analytic in D , which satisfy $f(0) = 0$. If $A_p(f)$ and $B_p(f)$ are two quantities associated with the functions $f \in \mathfrak{F}$, we write $A_p \approx B_p$ if these exist positive constants k_p and K_p , not depending on f , such that

$k_p A_p \leq B_p \leq K_p A_p$, $0 < p < \infty$. The relation \approx is clearly symmetric and transitive. An application of (5.1) to $\text{Re } W_t$, $t \geq 0$, gives

$$(5.2) \quad E v^{p/2} \approx E \text{Re } W_{\tau_D}^{*p} = E u(Z_{\tau_D})^{*p},$$

and similarly we have

$$(5.3) \quad E v^{p/2} \approx E v(Z_{\tau_D})^{*p},$$

which together give

$$(5.4) \quad E u(Z_{\tau_D})^{*p} \approx E v(Z_{\tau_D})^{*p}.$$

Now define, for a set $F \subset D$, the sets $A(F) = \{Z_t \in F \text{ some } t < \tau_D\}$ and $N_\sigma(F) = \{\theta: S_\sigma(\theta) \cap F \neq \emptyset\}$. It is proved in [12], in a slightly different form, that for each σ , $0 < \sigma < 1$, there are positive constants c_σ and C_σ such that, if F is a connected Borel set,

$$(5.5) \quad c_\sigma P(A(F)) \leq \ell(N_\sigma(F)) \leq C_\sigma P(A(F)).$$

(We may take C_σ^{-1} to be the $a(\sigma)$ of Theorem 4.5.) Note that if F is closed, $P(A(F))$ is the harmonic measure of F relative to $D - F$. An application of (5.5) to the sets $\{|u| > \lambda\}$ yields the beautiful inequality

$$(5.6) \quad c_\sigma P(u^* > \lambda) \leq \ell(N_\sigma(u) > \lambda) \leq C_\sigma P(u^* > \lambda), \quad \lambda > 0,$$

where $N_\sigma(u)(\theta) = \sup_{z \in S_\sigma(\theta)} |u(z)|$ is the nontangential maximal function of u . Integrating (5.6) times λ^{p-1} gives

$$(5.7) \quad E u^{*p} \approx \int_0^{2\pi} N_\sigma(u)(\theta)^p d\theta, \quad \sigma \text{ fixed},$$

and the equivalent result holds for v , which, when combined with (5.7), yields

$$(5.8) \quad \int_0^{2\pi} N_\sigma(u)^p(\theta) d\theta \approx \int_0^{2\pi} N_\sigma(v)^p(\theta) d\theta, \quad \sigma \text{ fixed}.$$

This is a major theorem of [12]. Together with previously existing results, (5.8) made it possible to prove

$$(5.9) \quad \int_0^{2\pi} N_\sigma(u)^p(\theta) d\theta \approx \|f\|_{H^p}^p = \lim_{r \rightarrow 1} (2\pi)^{-1} \int_0^{2\pi} |f(re^{i\theta})|^p d\theta, \quad \sigma \text{ fixed},$$

where $\|f\|_{H^p}$ is the H^p norm of f .

The relationship (5.9) can be used to give a number of quantities that are equivalent to the H^p norm for $f \in \mathfrak{F}$. Perhaps the most useful is to combine (5.9) with (5.2) and (5.7) to get

$$\|f\|_{H^p}^p \approx E v^{p/2}$$

In particular, $f \in H^p$ if and only if $E v^{p/2} < \infty$. For many purposes $E v^{p/2}$ is a simpler quantity to work with than $\|f\|_{H^p}$, as D. L.

Burkholder has recently shown. We illustrate by sketching a proof of the following theorem, first proved by Burkholder in [8], although implicit in the work of Lowell Hansen, [25], if f is univalent.

Theorem 5.1. Let f and g be analytic in D with $f(0)=g(0)=0$. Suppose $g \in H^p$ and that $\lim_{r \rightarrow 1} g(re^{i\theta})$ exists and belongs to $\mathfrak{C} - f(D)$ for almost every θ . Then $f \in H^p$.

The hypotheses of this theorem, together with equation (4.5) (Here F will be a segment $\{re^{i\theta}, R < r < 1\}$), imply there is a random variable $s < \tau_D$, such that

$$(5.10) \quad P(g(Z_s) \notin f(D)) = 1.$$

Let $\eta = \inf\{t > 0: W_t(g) \notin f(D)\}$. Then (5.10) implies $P(\eta < \nu(g)) = 1$. Let $\gamma = \inf\{t > 0: W_t(f) \notin f(D)\}$. Clearly $\gamma \geq \nu(f)$. Now γ and η have the same distribution, since both are the first exit time of a standard Brownian motion started at 0 from the set $f(D)$. Thus

$$E\nu(f)^{p/2} \leq E\gamma^{p/2} = E\eta^{p/2} \leq E\nu(g)^{p/2}.$$

Since $g \in H^p$, $E\nu(g)^{p/2} < \infty$, so $E\nu(f)^{p/2} < \infty$ and thus $f \in H^p$, completing the proof.

Albert Baernstein II has recently shown, without probability, that the hypothesis of Theorem 5.1 imply $\|f\|_{H^p} \leq \|g\|_{H^p}$. See [8]. For related recent applications of probability to analytic functions, also see [5], [7], [9], and [10].

6. Conjugate function inequalities. Let $f = u + iv$ be analytic in D , continuous in \bar{D} , and satisfy $v(0) = 0$. If U and \hat{V} denote, respectively, the restriction of u and v to the boundary of D , it is easily seen that U determines \hat{V} completely, and \hat{V} is called the conjugate function of U . This map can be extended so that the conjugate function of any totally finite signed measure is defined in a way compatible with the definition just given, but here we will work with U which are as just above, that is, boundary values of function analytic in D and continuous in \bar{D} . The collection of all functions on ∂D of this type will be denoted \mathcal{G} . In a number of senses, the distribution of \hat{V} cannot be too much

greater than that of U . An example is the following theorem of Kolmogorov. For a number of related theorems, see [35].

Theorem 6.1. For each p , $0 < p < 1$, there are positive constants K_p such that

$$(6.1) \quad \left(\int_0^{2\pi} |\tilde{U}(e^{i\theta})|^p d\theta \right)^{1/p} \leq K_p \int_0^{2\pi} |U(e^{i\theta})| d\theta, U \in \mathcal{U}.$$

Note that if we define

$$(6.2) \quad \sup_{\substack{U \neq 0 \\ U \in \mathcal{U}}} \frac{\left(\int_0^{2\pi} |\tilde{U}(e^{i\theta})|^p d\theta \right)^{1/p}}{\int_0^{2\pi} |U(e^{i\theta})| d\theta} = C_p,$$

then Theorem 6.1 is equivalent to $C_p < \infty$. We will sketch a proof of this under the simplifying assumption that $\int_0^{2\pi} U(e^{i\theta}) d\theta = 0$, which is of course equivalent to $f(0) = 0$. Let C_p^* be the supremum in (6.2) if \mathcal{U} is replaced by the class \mathcal{U}^* of those U in \mathcal{U} satisfying this condition. A proof that $C_p^* < \infty$ will be sketched.

Note that, since Z_{τ_D} is uniformly distributed on ∂D ,

$$(6.3) \quad \frac{1}{2\pi} \int_0^{2\pi} |U(e^{i\theta})|^p d\theta = E|v(Z_{\tau_D})|^p = E|\operatorname{Im} W_v|^p,$$

and

$$(6.4) \quad \frac{1}{2\pi} \int_0^{2\pi} |U(e^{i\theta})| d\theta = E|\operatorname{Re} W_v| = \lim_{t \rightarrow \infty} E|\operatorname{Re} W_v \wedge t|,$$

the last equality by the bounded convergence theorem. Also $W_0 = 0$, by our simplifying assumption.

Let \mathcal{M}_t be the class of all Markov times for W_t . In view of (6.3) and (6.4), if

$$\sup_{\substack{\eta \in \mathcal{M}_t \\ \eta \neq 0}} \frac{(E|\operatorname{Im} W_\eta|^p)^{1/p}}{\lim_{t \rightarrow \infty} E|\operatorname{Re} W_\eta \wedge t|} = C_p^*,$$

then $C_p' \geq C_p^*$, so if C_p' is finite then C_p^* is. A natural way to prove $C_p' < \infty$ is to find the $\eta = \mu$ which maximizes the ratio (6.5), and then to calculate

$$(6.6) \quad \frac{(E| \operatorname{Im} W_\mu |^p)^{1/p}}{\lim_{t \rightarrow \infty} E| \operatorname{Re} W_\mu \wedge t |}$$

which will of course be C_p' . A solution is $\mu = \inf\{t > 0: |\operatorname{Im} W_t| \geq 1, \operatorname{Re} W_t = 0\}$. This will not be proved here. It happens that C_p' is not only finite, but also $C_p' = C_p^*$. The examples necessary to show this are associated with the standard analytic function $g(z) = 2z/(1-z^2)$ mapping D into $\{z: |\operatorname{Im} z| \geq 1, \operatorname{Re} z = 0\}$, which is not surprising, since $v(g) = \mu$.

With a little more work it can be shown that $C_p' = C_p^*$ (see [17]), which is of course the best possible value for K_p in Theorem 6.1. The value is $C_p = (\frac{1}{2\pi} \int_0^{2\pi} |\operatorname{Im} g(re^{i\theta})|^p)^{1/p}$. This method is applied to other problems in [14] and [15].

7. Growth of analytic functions. If f is analytic in D , and $|f'(0)| = 1$, then $f(D)$ cannot be too small. There are several classical results which make this statement precise, and following is one way to do it probabilistically.

Theorem 7.1. There is a positive constant C such that if f is analytic in D then

$$P(v(f) > C|f'(0)|^2) > C.$$

This theorem will not be proved here. It is a consequence of the definition of v , the subharmonicity of $|f'(z)|^2$, and the lemma of [13].

Theorem 7.1 will first be used to prove Koebe's theorem, stated below. It is known that the largest possible value for the constant K is $1/4$, but we don't get this.

Theorem 7.2. There is a positive constant K such that, if f is analytic and univalent in D , and satisfies $f(0) = 0$ and $|f'(0)| = 1$, then $\{z: |z| < K\} \subset f(D)$.

Proof. Let us say that a curve $\gamma = \gamma_t$, $a \leq t \leq b$, encloses a set A if there are times $a \leq \alpha \leq \beta \leq b$ such that γ_t , $\alpha \leq t \leq \beta$, is a closed curve with none of A in the unbounded component of its complement, and define the radius of enclosure $R(\gamma)$ to be the largest r such that $\{z: |z| < r\}$ is enclosed by γ .

If there are numbers $a \leq t_1 \leq t_2 \leq b$ such that $|\gamma_{t_1}| = |\gamma_{t_2}|$ but $\text{Arg } \gamma_{t_1} \neq \text{Arg } \gamma_{t_2}$, it is easily seen that there exists an N such that $n \geq N$ implies γ_t^n , $a \leq t \leq b$, encloses 0 , since the changes in argument of γ_t^n are n times as great as those of γ_t . Thus, since almost every Brownian path has many such pairs of times, there is a random variable N , $P(N < \infty) = 1$, such that $n \geq N$ implies Z_t^n , $0 \leq t \leq \tau_D$, encloses $\{0\}$. Thus

$$(7.1) \quad \lim_{n \rightarrow \infty} P(Z_t^n, 0 \leq t \leq \tau_D, \text{ encloses } \{0\}) = 1.$$

But, since z^n is analytic, and maps ∂D onto itself, Z_t^n , $0 \leq t \leq \tau_D$, is just Brownian motion started at 0 up to the first time it leaves D . Thus each of the probabilities in (7.1) is the same, and

$$(7.2) \quad P(Z_t, 0 \leq t \leq \tau_D, \text{ encloses } \{0\}) = 1.$$

If $a > 0$, and $\tau_{aD} = \inf\{|Z_t| = a\}$, then $a^{-1}Z_t$, $0 \leq t \leq \tau_{aD}$, is Brownian motion up to the first time its absolute value is 1 , since

$a^{-1}z$ is analytic, so (7.2) implies $P(Z_t, 0 \leq t \leq \tau_{aD}, \text{encloses } \{0\}) = 1$. Since $P(\tau_{aD} \rightarrow 0) = 1$ as $a \rightarrow 0$, this gives

$$(7.3) \quad P(Z_t, 0 \leq t \leq \epsilon, \text{encloses } \{0\}) = 1, \text{ for each } \epsilon > 0.$$

Let K satisfy

$$(7.4) \quad P(R(Z_t, 0 \leq t \leq C) > K) > 1-C,$$

C as in Theorem 7.1. Such a K can be found since, by (7.3), if C is fixed and $K \rightarrow 0$ the probability in (7.4) approaches 1.

Applying (7.4) to W_t we get

$$\begin{aligned} P(R(W_t, 0 \leq t \leq v) > K) \\ &\geq P(R(W_t, 0 \leq t \leq C) > K \text{ and } v \geq C) \\ &\geq (1-C) - (1-C) > 0. \end{aligned}$$

Thus, since all the paths $W_t, 0 \leq t \leq v$, lie in $f(D)$, $f(D)$ contains at least one closed curve enclosing $\{|z| < K\}$, which, since $f(D)$ is simply connected (this is the only place univalence is used), proves Theorem 7.2.

To conclude this section the following theorem will be proved.

Theorem 7.3. Let $h(z)$ be an entire function satisfying
 $\lim_{z \rightarrow \infty} |h(z)| = 0$. Then either $h(z) = +1$ for infinitely many z , or
 $h(z) = -1$ for infinitely many z , or both.

This is equivalent to Picard's big theorem, which says that an entire function $g(z)$ which is not a polynomial must satisfy $g(z) = a$ for infinitely many z , with at most one exceptional value a . The classical result, that if $g(z)$ is not a polynomial then $\lim_{z \rightarrow \infty} |g(z)| = 0$, must be used to get this equivalence.

First several lemmas will be stated. The first of these is an extension of Theorem 7.1, and is a little more general than equation 10 of [13]. It can be proved in a similar way.

Let rD be the disc of radius r around 0, let $\tau(r) = \inf\{t: |Z_t| = r\}$, (so that $\tau(1) = \tau_D$), let f be analytic in rD , define ρ as in (2.3), and let $M(r) = \max_{|z| \leq r} |f'(z)|$. Then

Lemma 7.3. For each s , $0 < s < 1$, there is a positive constant $C(s)$, which decreases as s increases, such that

$$P(\rho(\tau(\alpha))) > C(s)M(s\alpha)^2\alpha^2 > C(s).$$

The statement of the next lemma is related to Theorem 3.1. Although Z_t started at 0 in that theorem this was done only to simplify notation, and in fact Z_t could have started at any point in $\hat{0}$. The proof of the following lemma follows from the proof of Theorem 3.1 of [16].

Lemma 7.4. If $Z_0 = z \in \hat{0}$, there is a time τ , $P_z(\tau < \infty) = 1$, such that $s > \tau$ and $Z_s \in \hat{0}$ implies Z_t , $0 \leq t \leq s$, is not homotopic to 0 in $\hat{\phi}$. Furthermore, there is a function $\lambda(x)$, which does not depend on $z \in \hat{0}$, satisfying $\lambda(x) \rightarrow 1$ as $x \rightarrow \infty$, such that

$$P_z(\tau \leq x) > \lambda(x).$$

Next the following weak form of the Picard-Schottsky Theorem (see [1]) is proved.

Lemma 7.5. For each r , $0 < r < 1$, there is a positive constant $K(r)$, such that $M(r) > K(r)$ and $|f(0)| \leq \frac{1}{20}$ implies $f(D)$ contains either +1 or -1 or both.

Proof. Let r be fixed. Either Lemma 7.3 or Schwartz's lemma guarantees that, if $M(r)$ is large enough, then $f(D_{\sqrt{r}})$ is not contained in $\hat{0}$, and we assume that $M(r)$ is this large. Let D_x be the

largest disc with center 0 such that $f(D_x) \subset \hat{0}$.

Let $T = \inf\{t > \tau(\sqrt{x}) : Z_t = 1 \text{ or } x\}$, and let $S = \inf\{t > \tau(r^{1/4}) : Z_t = 1 \text{ or } x\}$. The Strong Markov Property and an equation like (2.2) give

$$(7.5) \quad P(|Z_T| = x | Z_t, 0 \leq t \leq \tau(\sqrt{x})) = 1/2.$$

Since $x < \sqrt{r}$, $\sqrt{x} < \sqrt[4]{r}$, so $C(\sqrt{x}) > C(\sqrt[4]{r})$, and Lemma 7.3 with $s = \alpha = \sqrt{x}$ implies

$$(7.6) \quad P(\rho(\tau(\sqrt{x})) > C(r^{1/4}) M(x)^2 x) > C(r^{1/4}).$$

By the mean value theorem, $xM(x) \geq (1/10) - (1/20)$, so, since $\rho(T) \geq \rho(\tau(\sqrt{x}))$, (7.5) and (7.6) give

$$(7.7) \quad P(\rho(T) > C(r^{1/4}) (400x)^{-1} \text{ and } Z_T \in D_x) > C(r^{1/4})/2.$$

Also,

$$(7.8) \quad P(|Z_S| = x | Z_t, 0 \leq t \leq \tau(r^{1/4})) = \ln(r)/4\ln(x),$$

and an argument similar to the one which gave (7.7) gives

$$(7.9) \quad P(\rho(S) > C(r^{3/4}) M(r)^2 \sqrt{r} \text{ and } Z_S \in D_x) > C(r^{3/4}) \ln(r)/4\ln(x).$$

From (7.7) and (7.9) it can be deduced that, if $M(r)$ is large enough, then either

$$(7.10) \quad P(\rho(S) > y \text{ and } Z_S \in D_x) > 1 - \lambda(y), \text{ for some } y,$$

or

$$(7.11) \quad P(\rho(T) > y \text{ and } Z_T \in D_x) > 1 - \lambda(y), \text{ for some } y,$$

where $\lambda(y)$ is as in Lemma 7.4. (If x is very small, (7.10) holds, while if x and $M(r)$ are large then (7.11) will hold.)

Now (7.10) guarantees $P(\rho(S) > \tau, Z_S \in D_x) > 0$, and (7.11) guarantees $P(\rho(T) > \tau, Z_T \in D_x) > 0$, τ as in Lemma 7.4, so, if $M(r)$ is large enough, either $+1$ or -1 or both are in $f(D)$, by the argument of Section 3.

Now let Γ be the region $\{z: \frac{1}{2} < |z| < 2 \text{ and } -\pi/4 < \text{Arg } z < 3\pi/2\}$.

The following lemma can be proved in a manner similar to the proof of the last lemma, or can be proved using this lemma and the existence of a univalent analytic function mapping Γ onto D .

Lemma 7.6. Let g be analytic in Γ and let $|g(1)| \leq 1/20$. Let $\Gamma_\epsilon = \{z \in \Gamma: \text{distance}(z, \partial\Gamma) > \epsilon\}$. There is a constant $\Theta(\epsilon) > 0$, such that $|g'(\epsilon)| > \Theta(\epsilon)$ for $z \in \Gamma_\epsilon$ implies that $g(\Gamma)$ contains either $+1$ or -1 or both.

Now the proof of Theorem 7.3 can be completed. Let z_1 and z_2 be two points on $\{z: |z| = R\}$, $R \geq 1$, such that $|h(z_1)| \leq 1/20$ and $|h'(z_2)| > \Theta(.01)$. Such z_1, z_2 occur for arbitrarily large R . Both $h(z_1/z) = g_1(z)$ and $h(z_1/z) = g_2(z)$ are analytic in Γ and satisfy $|g_i(1)| \leq 1/20$. Furthermore, $|g'_1(z_2/z_1)| = R|f'(z_2)| = |g'_2(z_1/z_2)|$, and one of $z_1/z_2, z_2/z_1$ is in Γ and at least a distance of .01 from the boundary of Γ . Thus Lemma 7.6 guarantees that one of $g_1(\Gamma), g_2(\Gamma)$ contains either -1 or 1 , which implies that h takes on either -1 or 1 in $R/2 < |z| < 2R$. Since this is true for arbitrarily large R , h takes on either -1 or 1 infinitely often.

8. Concluding remarks. As was mentioned in the introduction, some probabilistic expressions have arisen in the preceding sections which are directly translatable to more conventional expressions, for example,

$E|W_v| = \int_0^{2\pi} |f(e^{i\theta})| d\theta/2\pi$, while others cannot be so handled. The most interesting of these, to me, is the time $v(f)$. The distribution of $v(f)$ is an intuitively appealing measure of the size of $f(D)$, since if f is univalent then v is the first exit time of standard Brownian motion from $f(D)$ and, even if f is not univalent, a similar interpretation can be given using Riemann surfaces. In the preceding sections there are a number of instances where information about the size of $v(f)$ translates easily into more conventional information about the size of $f(D)$. However, little of a precise nature is known about v . For example, it is not even known in what sense, with regard to $v(f)$, z is the smallest analytic function in D satisfying $|f'(0)| = 1$. It would be nice if

$$(8.1) \quad P(v(z) \geq \lambda) \leq P(v(f) \geq \lambda), \quad \lambda > 0,$$

for all such f , which would improve Theorem 7.1, and it is probably true that

$$(8.2) \quad E\phi(v(z)) \leq E\phi(v(f)),$$

for all increasing and convex functions ϕ on $[0, \infty)$, but neither of these results is known (although it is known that (8.2) holds if $\phi(x) = x$; see equation (2) on page 309 of [27]). In the other direction, it can be asked in what sense, with regard to v , the Koebe function $\frac{z}{(1-z)^2}$ is the largest univalent function in D such that $|f'(0)| = 1$.

The reader interested in these questions should see Burkholder, [7].

The subject of Brownian motion on Riemann surfaces is outside the scope of this paper, but it has been successfully defined, and, since the analogue of Levy's theorem holds, can be used to study analytic functions on Riemann surfaces. Many of the arguments in Section 4 were adapted from proofs for such functions. Now on some surfaces, for example the plane, Brownian motion is recurrent, that is returns to each neighborhood of its starting point at arbitrarily large times, while Brownian motion on other surfaces, for example an open disc, does not have this property. If there is an analytic function mapping one Riemann surface onto another, then, by Levy's theorem, Brownian motion is recurrent on both or neither. Kakutani used such ideas to study when there exist such mappings, in [29]. See McKean's book, [32], for more about Brownian motion on Riemann surfaces. It is in this context that a number of interesting applications of function theory to Brownian motion occur.

The whole subject of conformal invariance has been studied in a more abstract setting, in [29]. Finally, we remark that there are uses of Brownian motion in analysis other than those mentioned here. See, for examples, J. P. Kahane's paper [28].

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