# Hausdorff dimension of the boundary of Brownian bubbles 

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Based on joint work with:
T. Mountford (EPF - Lausanne)

## The Brownian sheet

A standard two-parameter Brownian sheet is a centered Gaussian random field $W=\left(W\left(t_{1}, t_{2}\right),\left(t_{1}, t_{2}\right) \in \mathbb{R}_{+}^{2}\right)$ defined on a probability space $(\Omega, \mathcal{F}, P)$, with continuous sample paths and covariance

$$
E\left[W\left(s_{1}, s_{2}\right) W\left(t_{1}, t_{2}\right)\right]=\min \left(s_{1}, t_{1}\right) \min \left(s_{2}, t_{2}\right)
$$

For fixed $t_{2}, t_{1} \mapsto W\left(t_{1}, t_{2}\right)$ is a Brownian motion (with speed $t_{2}$ ).
References:
1970's: L. Pitt, S. Orey \& W. Pruitt, R. Pyke, R.J. Adler
1980's: W. Kendall, J.B. Walsh, D. Nualart
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Issues: Sample path properties, Markov properties, potential theory, level sets, small ball probabilities, hitting probabilities, multiple points.

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## Level sets and bubbles

For $x \in \mathbb{R}$, the level set of $W$ at level $x$ is the random closed set

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L(x):=\left\{\left(t_{1}, t_{2}\right) \in \mathbb{R}_{+}^{2}: W\left(t_{1}, t_{2}\right)=x\right\} .
$$

The complement of the level set is the union of two random open sets

$$
\begin{aligned}
& L_{+}(x):=\left\{\left(t_{1}, t_{2}\right) \in \mathbb{R}_{+}^{2}: W\left(t_{1}, t_{2}\right)>x\right\} \\
& L_{-}(x):=\left\{\left(t_{1}, t_{2}\right) \in \mathbb{R}_{+}^{2}: W\left(t_{1}, t_{2}\right)<x\right\}
\end{aligned}
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Definition. A Brownian bubble is one connected component of $L_{+}(x)$ or $L_{-}(x)$.
(Recall that any open subset of $\mathbb{R}_{+}^{2}$ is a countable disjoint union of connected components.)

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## Hausdorff dimension of level sets

## Theorem 1 (R.J. Adler, 1978)

A.s., for all $x \in \mathbb{R}, \operatorname{dim}_{\mathcal{H}} L(x)=1.5$

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Theorem 2 (T. Mountford, 1993)
Fix }x\in\mathbb{R}\mathrm{ . As, the Hausdorff dimension of the boundary of any Brownian
bubble is: }\geqslant1.25\mathrm{ and <1.5
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Interpretation: "Most of $L(x)$ is not part of the boundary of any bubble." Comparison with standard Brownian motion:
bubbles $\longleftrightarrow$ excursions above/below level $x$; boundaries of bubbles $\longleftrightarrow$ extremities of excursion intervals.

There are countably many extremities of excursion intervals (dimension 0), but the dimension of level sets of standard Brownian motion is $\frac{1}{2}$
Question. Do all bubble boundaries have the same dimension? If so, what is it?

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## Explanation for Adler's theorem

Upper bounds on Hausdorff dimension $\longleftarrow$ coverings.
Let

$$
V_{n}:=\left\{\left(1+i 2^{-2 n}, 1+j 2^{-2 n}\right): i, j \in\left\{0, \ldots, 2^{2 n}-1\right\} .\right.
$$

Then $V_{n}=$ vertices of a grid in $[1,2]^{2}, \quad \sharp V_{n}=2^{4 n}$.
For $t \in \mathbb{E}_{n}$, define $E_{n}(t):=$ the square in the grid with lower left corner at $t$. P2 One covering of $L(x) \cap[1,2]^{2}$, with diameter $c 2^{-2 n}$, is:

$$
\left\{E_{n}(t): t \in V_{n}, E_{n}(t) \cap L(x) \neq \emptyset\right\}
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Calculation:


Now

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P\left\{E_{n}(t) \cap L(x) \neq \emptyset\right\} \simeq P\left\{|W(t)-x| \leqslant 2^{-n}\right\} \simeq 2^{-n}
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Calculation:

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E\left[\sum_{t \in V_{n}}\left(2^{-2 n}\right)^{\alpha} 1_{\left\{E_{n}(t) \cap L(x) \neq \emptyset\right\}}\right]=\left(2^{-2 n}\right)^{\alpha}\left(2^{2 n}\right)^{2} P\left\{E_{n}(t) \cap L(x) \neq \emptyset\right\}
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$$
\leqslant 2^{(4-2 \alpha) n} 2^{-n}=2^{(3-2 \alpha) n} \rightarrow 0
$$

as $n \rightarrow \infty$ if and only if $\alpha>\frac{3}{2}$.

## Towards the dimension of bubble boundaries

Let $\mathcal{C}_{1}$ be a bubble of height $\geqslant 1$ (in $[1,2]^{2}$ ). Then:
$t \in \partial \mathcal{C}_{1} \Longleftrightarrow W(t)=x$ and for all $\varepsilon>0$, there exists a path $\Gamma$ with

$$
d(\Gamma(0), t)>\varepsilon \text { and } W(\Gamma(\cdot))-x \text { hits } 1 \text { before } 0 .
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## P3

Covering of $\partial \mathcal{C}_{1} \cap[1,2]^{2}$

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\left\{E_{n}(t): E_{n}(t) \cap L(x) \neq \emptyset \text { and } F(t) \text { occurs }\right\}
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where

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F(t)=\{\exists \Gamma: \Gamma(0)=t \text { and } W(\Gamma(\cdot))-x \text { hits } 1 \text { before } 0\}
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Should examine the behavior as $n \rightarrow \infty$ of

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Main difficulty in estimating $P\left\{F(t)\left||W /(t)-x|<2^{-n\}}\right.\right.$ : there are infinitely many possible paths, and these can be arbitrarily "twisty" [D. \& Walsh, 1993]

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\begin{aligned}
& \sum_{t \in V_{n}}\left(2^{-2 n}\right)^{\alpha} P\left\{|W(t)-x| \leqslant 2^{-n}\right\} P\left\{F(t)| | W(t)-x \mid \leqslant 2^{-n}\right\} \\
& \quad \simeq 2^{4 n} 2^{-2 \alpha n} 2^{-n} P\left\{F(t)| | W(t)-x \mid \leqslant 2^{-n}\right\}
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Main difficulty in estimating $P\left\{F(t)\left||W(t)-x| \leqslant 2^{-n}\right\}\right.$ : there are infinitely many possible paths, and these can be arbitrarily "twisty" [D. \& Walsh, 1993].

## Local decomposition of the Brownian sheet

The event $F(t)$ is "local": either 0 is hit rather quickly, or not, and in this case, $W-x$ will typically escape to a height of order 1 (the same occurs for Brownian motion).

Local decomposition of $W$ [W. Kendall, 1980]: Fix $t=\left(t_{1}, t_{2}\right)$. For $u_{1}, u_{2} \in \mathbb{R}$,

$$
W\left(t_{1}+u_{1}, t_{2}+u_{2}\right)=W\left(t_{1}, t_{2}\right)+B_{1}^{t}\left(u_{1}\right)+B_{2}^{t}\left(u_{2}\right)+\mathcal{E}^{t}\left(u_{1}, u_{2}\right)
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where:

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\begin{aligned}
& B_{1}^{t}, B_{2}^{t} \text { are independent (two-sided) BM's, and } \\
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This suggest to study additive Brownian motion

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## Gambler's ruin problem for additive BM

Let $X=\left(X\left(u_{1}, u_{2}\right),\left(u_{1}, u_{2}\right) \in \mathbb{R}^{2}\right)$ be an additive Brownian motion.
For $x \in[0,1]$, define

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\mathbb{E}(x):=P\{\exists \text { path } \Gamma: \Gamma(0)=(0,0), X(\Gamma(\cdot)) \text { hits } 1 \text { before } 0 \mid X(0,0)=x\}
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Problem. Estimate $\mathbb{E}(x)$.
P4
Main difficulty: there is no constraint on the path $\Gamma$ : one has to consider all paths, with no restrictions.

Related problem. For $X(0,0) \neq 0$, let $\mathcal{C}_{(0,0)}$ be the bubble "stradling" $(0,0)$
Question. For $a>0$, what is the probability that the bubble $\mathcal{C}_{(0,0)}$ extends at least $a$ units away from the origin?

That is, estimate


By scaling, $\mathbb{D}(x, a)=\mathbb{D}(x / \sqrt{a}, 1)$, and we expect $\mathbb{D}(x, 1) \simeq \mathbb{E}(x)$ for $x \downarrow 0$

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## Gambler's ruin

## Theorem 3 (D. \& Mountford)

For $x \in[0,1]$,

$$
\mathbb{E}(x)=\alpha_{1} x^{\lambda_{1}}+\alpha_{2} x^{\lambda_{2}}+\alpha_{3} x^{\lambda_{3}}+\alpha_{4} x^{\lambda_{4}}
$$

where

$$
\left\{\lambda_{1}, \lambda_{2}, \lambda_{3}, \lambda_{4}\right\}=\left\{\frac{1}{2}(5 \pm \sqrt{13 \pm 4 \sqrt{5}})\right\}
$$

$\lambda_{1}=\frac{1}{2}(5-\sqrt{13+4 \sqrt{5}}) \simeq 0.158<\lambda_{2} \simeq 1.49<\cdots$
$\alpha_{1} \simeq 0.939, \alpha_{2}=\ldots$ (exact, explicit formulas are given).
In particular, $\mathbb{E}(x) \simeq x^{\lambda_{1}}$ as $x \downarrow 0$.
Comparison. For standard BM , we would have $\mathbb{E}(x) \simeq x \ll x^{\lambda_{1}}$.
Theorem 3 is somewhat surprising!

## Escape probabilities

## Corollary 1

There exist $0<c<C<\infty$ such that, for all $a \geqslant x^{2}$,

$$
c\left(\frac{x}{\sqrt{a}}\right)^{\lambda_{1}} \leqslant \mathbb{D}(x, a) \leqslant C\left(\frac{x}{\sqrt{a}}\right)^{\lambda_{1}}
$$

Proving Corollary 1 from Theorem 3 requires some effort.

## Main result

## Theorem 4 (D. \& Mountford)

Fix $x \in \mathbb{R}$. For the Brownian sheet, the Hausdorff dimension of the boundary of every $x$-bubble is

$$
\frac{3}{2}-\frac{\lambda_{1}}{2}=\frac{1}{4}(1+\sqrt{13+4 \sqrt{5}}) \simeq 1.421
$$

Once Theorem 3 and Corollary 1 are proved, the road map to prove Theorem 4 is fairly clear. Carrying out these steps requires some effort.
Will explain why Theorem 3 is true, then give some ideas on how to deduce Theorem 4

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## Proving Theorem 3 (gambler's ruin probabilities for ABM)

## Theorem 5 (D. \& Walsh, 1993)

There is a specific path $\Gamma^{\circ}$ such that

$$
\mathbb{E}(x)=P\left\{X\left(\Gamma^{\circ}(\cdot)\right) \text { hits } 1 \text { before } 0 \mid X(0,0)=x\right\}
$$

P6 Explain construction of $\Gamma^{0}$ : the DW-algorithm.

## Lemma

The sequence $M_{0}=x, M_{1}, M_{2}, \ldots$ of successive maxima encountered along the horizontal/vertical segments of the path $\Gamma^{\circ}$ is Markov of order 2, with transition probabilities

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$$
P\left\{M_{n+1} \in d z \mid M_{n}=y, M_{n-1}=x\right\}=f(x, y, z) d z, \quad z>y>x
$$

where

$$
f(x, y, z)=\frac{2(y-x)}{z^{2}}-\frac{2(y-x)^{2}}{z^{3}}
$$

and

$$
P\left\{M_{n+1}=y \mid M_{n}=y, M_{n-1}=x\right\}=\left(\frac{x}{y}\right)^{2}
$$

## Study of the Markov chain $\Theta_{n}=\left(M_{n-1}, M_{n}\right)$

State space: $\mathcal{S}=\left\{\left(y_{1}, y_{2}\right) \in \mathbb{R}_{+}^{2}: 0<y_{1} \leqslant y_{2}\right\}$
P7 Consider the paths of $\left(\Theta_{n}\right)$
Define the subsets:

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\begin{aligned}
& \text { WIN }:=\left\{\left(y_{1}, y_{2}\right) \in \mathcal{S}: y_{2} \geqslant 1\right\} \\
& \text { LOSE }:=\left\{\left(y_{1}, y_{2}\right) \in \mathcal{S}: y_{2}=y_{1}\right\} .
\end{aligned}
$$

P8 and set

$$
\alpha(x, y)=P\left\{\left(\Theta_{n}\right) \text { visits LOSE before WIN } \mid \Theta_{1}=(x, y)\right\}
$$

Then

$$
\alpha(x, y)=\left(\frac{x}{y}\right)^{2}+\int_{y}^{1} d z f(x, y, z) \alpha(y, z)
$$

This is an unusual sort of linear integral equation (but similar to the system of equations for absorption probabilities for Markov chains). After several manipulations, one checks that:

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$$
\alpha(x, y)=P\left\{\left(\Theta_{n}\right) \text { visits LOSE before WIN } \mid \Theta_{1}=(x, y)\right\}
$$

Then

$$
\begin{equation*}
\alpha(x, y)=\left(\frac{x}{y}\right)^{2}+\int_{y}^{1} d z f(x, y, z) \alpha(y, z) \tag{1}
\end{equation*}
$$

This is an unusual sort of linear integral equation (but similar to the system of equations for absorption probabilities for Markov chains). After several manipulations, one checks that:

## Solving the integral equation

Solving (1) is equivalent to soving the linear system of o.d.e.'s

$$
\underline{\dot{x}}(y)=A \cdot \underline{x}(y)+\underline{b}, \quad y>0,
$$

where $A$ is the $6 \times 6$ matrix and $\underline{b}$ and $\underline{x}(0)$ are the column vectors
$A=\left(\begin{array}{rrrrrr}0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & -9 & 6 & 4 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ -8 & 2 & 0 & 28 & -26 & 9\end{array}\right) \quad \underline{b}=\left(\begin{array}{r}0 \\ 0 \\ -2 \\ 0 \\ 0 \\ -6\end{array}\right) \quad \underline{x}(0)=\left(\begin{array}{r}0 \\ -1 \\ -3 \\ 0 \\ 1 \\ -4\end{array}\right)$

This yields an explicit formula for $\alpha(x, y)$, via the 4 real eigenvalues $\lambda_{1}, \ldots, \lambda_{4}$ and eigenvectors of A. Finally,

$$
\mathbb{E}(x)=1-E\left[\alpha\left(x, H_{1}\right)\right],
$$

where $P\left\{H_{1} \leqslant y\right\}=\left(P_{x}\{B(\cdot) \text { hits } 0 \text { before } y\}\right)^{2}=\left(\frac{y-x}{y}\right)^{2}$
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This leads to the explicit formula for $\mathbb{E}(x)$.

## Proving Theorem 4

Theorem. $\operatorname{dim}_{H}$ "bubble" $=\frac{3}{2}-\frac{\lambda_{1}}{2}$.
Part 1. Upper bound: $\operatorname{dim}_{H}$ "bubble" $\leqslant \frac{3}{2}-\frac{\lambda_{1}}{2}$.
Use the covering argument discussed previously:

$$
\begin{align*}
\sum_{t \in V_{n}} & \left(2^{-2 n}\right)^{\alpha} P\left\{|W(t)-x| \leqslant 2^{-n}\right\} P\left\{F(t)| | W(t)-x \mid \leqslant 2^{-n}\right\} \\
& \simeq 2^{4 n} 2^{-2 \alpha n} 2^{-n} P\left\{F(t)| | W(t)-x \mid \leqslant 2^{-n}\right\} \\
& \simeq 2^{(3-2 \alpha) n}\left(2^{-n}\right)^{\lambda_{1}}  \tag{2}\\
& =2^{\left(3-2 \alpha-\lambda_{1}\right) n} \\
& \longrightarrow 0 \quad \text { if } \alpha>\frac{3-\lambda_{1}}{2}
\end{align*}
$$

Note. (2) concerns the Brownian sheet, not $A B M$ : some effort is needed to go from one to the other ("robustness" of the DW-algorithm).

## Proving Theorem 4

Part 2. Lower bound: $\operatorname{dim}_{H}$ "bubble" $\geqslant \frac{3}{2}-\frac{\lambda_{1}}{2}$.
Energy method: For $\alpha<\frac{3}{2}-\frac{\lambda_{1}}{2}$, seek a measure $\mu$ supported on the boundary of a bubble, such that

$$
\iint \frac{\mu(d s) \mu(d t)}{|t-\boldsymbol{s}|^{\alpha}}<\infty
$$

Via a "second moment argument", the key estimate is:

## Part 2 (continued)

## Lemma

For $s, t \in[1,2]^{2}$, with $\left|s_{1}-t_{1}\right| \simeq 2^{2(k-n)},\left|s_{2}-t_{2}\right| \simeq 2^{2(\ell-n)}(1 \leqslant k<\ell \leqslant n)$,

$$
P\left\{|W(t)| \leqslant 2^{-n}, F(t),|W(s)| \leqslant 2^{-n}, F(s)\right\} \leqslant 2^{-n} 2^{-\ell}\left(2^{-k \lambda_{1}}\right)^{2}\left(2^{\ell-n}\right)^{\lambda_{1}}
$$

(recall that $F(t)=\{\exists \Gamma: \Gamma(0)=t$ and $W(\Gamma(\cdot))$ hits 1 before 0$\}$; here $x=0$.)
Explanation of each factor:
$W(t) \simeq 2^{-n}:$ prob.$\simeq 2^{-n}$
$W(s) \simeq 2^{-n}\left(\right.$ given $\left.W(t) \simeq 2^{-n}\right):$ prob. $\simeq \frac{2^{-n}}{2^{\ell-n}}=2^{-\ell}$
$F(t) \cap F(s)$ : first both paths reach level $2^{k-n}$ units: prob. $\left[\left(2^{-k}\right)^{\lambda_{1}}\right]^{2}$
In the big rectangle, the maximum of $W$ is $\simeq 2^{\ell-n}$. Starting from this level, one path (at least) must reach level 1 before 0: prob. $\simeq\left(2^{\ell-n}\right)^{\lambda_{1}}$

A good bound is obtained by multiplying these factors (even though the events are not independent!)

## Part 2 (continued)

## Lemma

For $s, t \in[1,2]^{2}$, with $\left|s_{1}-t_{1}\right| \simeq 2^{2(k-n)},\left|s_{2}-t_{2}\right| \simeq 2^{2(\ell-n)}(1 \leqslant k<\ell \leqslant n)$,

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$$
\begin{aligned}
& W(t) \simeq 2^{-n}: \text { prob. } \simeq 2^{-n} \\
& W(s) \simeq 2^{-n}\left(\text { given } W(t) \simeq 2^{-n}\right): \text { prob. } \simeq \frac{2^{-n}}{2^{\ell-n}}=2^{-\ell} . \\
& F(t) \cap F(s): \text { first both paths reach level } 2^{k-n} \text { units: prob. }\left[\left(2^{-k}\right)^{\lambda_{1}}\right]^{2} .
\end{aligned}
$$

In the big rectangle, the maximum of $W$ is $\simeq 2^{\ell-n}$. Starting from this level, one path (at least) must reach level 1 before 0: prob. $\simeq\left(2^{\ell-n}\right)^{\lambda_{1}}$.
A good bound is obtained by multiplying these factors (even though the events are not independent!).

## References

Mountford, T.S. (1993). Estimates of the Hausdorff dimension of the boundary of positive Brownian sheet components. Sém. de Probabilités XXVII Lect. Notes in Math. vol. 1557, pp.233-255. Springer.

Dalang, R.C. \& Walsh, J.B. (1993) The structure of a Brownian bubble. Probab. Th. Relat. Fields 96, 475-501.

Mörters, P. \& Peres, Y. (2010). Brownian motion. Cambridge University Press.
Dalang, R.C. \& Mountford, T. (2017). Hausdorff dimension of the boundary of bubbles of additive Brownian motion and of the Brownian sheet (145pp). http://arxiv.org/abs/1702.08183

