

# Random measures on planar curves

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Two theorems in this paper investigate Fourier—Stieltjes transforms of measures on curves in the plane; the first is more general in respect of the curves investigated and more precise in the inequalities obtained for Fourier transforms, while the second introduces a more powerful and flexible technique. The first is proved by elementary methods, insofar as it avoids the theory of stochastic processes and martingales.

A survey of Gaussian processes in related extremal problems is given by Kahane [3, ch. 13—15], who extends some results to stable processes in [4]. A set of the type  $S$  occurring in the first theorem is often called a *Salem set* of dimension  $\beta$  [8], [5, VIII]. Non-linear transformations and Hausdorff dimension are studied in [6].

1. Let  $\Gamma$  be a curve of class  $C^2$  in the plane  $\mathbf{R}^2$ , of positive curvature. An elementary proof is provided for the following *theorem*:

*For each  $\beta$  in  $(0, 1)$  there exists a compact set  $S \subseteq \Gamma$ , of Hausdorff dimension  $\beta$ , and a positive measure  $\mu$  on  $S$ , such that*

$$|\hat{\mu}(u)| = o(1) \|u\|^{-\beta/2} \quad \text{for } u \text{ in } \mathbf{R}^2.$$

For definiteness we suppose that  $\Gamma$  is described by co-ordinates  $-2 \leq x \leq 2$ , and  $y = y(x)$ , where  $y$  is of class  $C^2$  and  $y'' > 0$ . Whenever  $\lambda$  is a measure on  $[-1, 1]$  we denote by  $\mu$  its transform by the mapping of  $[-2, 2]$  onto  $\Gamma$ . For present purposes,  $\mu$  is best defined by its Fourier—Stieltjes transform on  $\mathbf{R}^2$ :

$$\hat{\mu}(u_1, u_2) = \int \exp(-iu_1x - iu_2y(x)) \lambda(dx), \quad u \in \mathbf{R}^2.$$

**Lemma.** *Let  $f$  be an element of  $C^2[-2, 2]$  and  $f=0$  outside  $(-1, 1)$ . For any  $\varepsilon > 0$  and integer  $N > N_\varepsilon$  there is an element  $g_1 \cong 0$  of  $C^2[-2, 2]$  so that*

- (i) *The closed support of  $g_1$  is covered by  $10N^\beta \log^2 N$  intervals of length  $N^{-1}$ .*
- (ii) *The measures  $\mu$  and  $\mu_1$  corresponding to  $f(x)dx$  and  $g_1(x)f(x)dx$  fulfil the inequality*

$$|\hat{\mu}(u) - \hat{\mu}_1(u)| \leq \varepsilon(1 + \|u\|)^{-\beta/2}, \quad u \in \mathbf{R}^2.$$

In the proof we use a special partition  $1 = \sum h_k$ , in which  $0 \leq h_k \leq 1$ ,  $h_k \in C^2(-\infty, \infty)$ ,  $|h'_k| \leq CN$  and  $h_k = 0$  outside  $((k-1)/2N, (k+1)/2N)$ . The existence of a partition  $1 = \sum H_k$ , adapted to  $N=1$ , is well known, and we have only to choose  $h_k(x) \equiv H_k(Nx)$ . In the construction of  $g_1 = \sum \xi_k h_k$ , with certain random variables  $\xi_k$ , we need bounds for the partial integrals

$$b_k(u) = \int h_k(x) f(x) \exp(-iu_1 x - iu_2 y(x)) dx.$$

In fact we have the inequalities

- (a)  $|b_k(u)| \leq CN^{-1}$  for all  $u$ .
- (b)  $\sum |b_k(u)|^2 \leq C \|u\|^{-1}$  when  $\|u\| \leq N$ .

Here  $C$  depends on  $f$  and  $y(x)$  but not on  $N$ ; it is easy to see that  $b_k(u) = 0$  if  $|k| > 2N$ , and (a) is a consequence of the properties of  $h_k$ . As for (b), we can easily dispose of the special case  $|u_1| \geq 2A|u_2|$ , where  $A = \sup \{|y'(x)|\}$ . For the analysis leading to van der Corput's inequality [9, p. 116] yields  $|b_k(u)| \leq C \|u\|^{-1}$ , and then  $\sum |b_k(u)|^2 \leq C \|u\|^{-2} N \leq C \|u\|^{-1}$ . Henceforth, in the proof of (b) we assume  $|u_1| \leq 2A|u_2|$ .

We let  $m_k$  be the minimum of  $|u_1 + u_2 y'(x)|$  on the interval  $((k-1)/2N, (k+1)/2N)$ , and obtain  $|b_k(u)| \leq C m_k^{-1}$  by the argument mentioned before. We also have the bounds  $|b_k(u)| \leq CN^{-1}$  and  $|b_k(u)| \leq C'|u_2|^{-1/2} \leq C \|u\|^{-1/2}$  by [9, p. 116]. Suppose now that  $m_k$  attains a minimum at  $k=p$ ; then  $|m_{k+p}| \leq C' p |u_2| N^{-1} \leq Cp \|u\| N^{-1}$  if (say)  $|p| \geq 3$ . When  $N \leq \|u\| \leq N^2$  we use the bound  $|b_{k+p}(u)| \leq CN^{-1}$  for the range  $|p| \leq 3 \|u\|^{-1} N^2$ , and the lower bound on  $m_k$  for other values of  $p$ . When  $N^2 \geq \|u\|$  we use the bound  $|b_k(u)| \leq C \|u\|^{-1/2}$  for  $|p| \leq 3N \|u\|^{-1/2}$ , and  $C m_k^{-1}$  outside.

To construct  $g_1$  from all these preliminaries we set  $p_N = N^{\beta-1} \log^2 N$ , and take independent random variables  $\xi_k$  with law

$$P(\xi_k = p_N^{-1}) = p_N, \quad P(\xi_k = 0) = 1 - p_N.$$

Then  $g_1 = \sum \xi_k h_k$ .

For each fixed  $u$  in  $\mathbf{R}^2$ ,  $\hat{\mu}_1(u) - \hat{\mu}(u) = \sum (\xi_k - 1) b_k(u)$ , a sum of independent terms of magnitude  $CN^{-1} p_N^{-1}$  and total variance  $Cp_N^{-1} \sum |b_k^2(u)|$ . Classical bounds for expected values of  $\exp(t \operatorname{Re} \hat{\mu}_1 - \operatorname{Re} \hat{\mu})$  and  $\exp(t \operatorname{Im} \hat{\mu} - \operatorname{Im} \hat{\mu}_1)$  are valid provided  $|t| \leq C^{-1} N p_N$ . Choosing then  $t = \eta N^{1-\beta/2} p_N$  for a small constant  $\eta > 0$  we conclude

$$P(|\hat{\mu}_1(u) - \hat{\mu}(u)| \geq \varepsilon N^{-\beta/2}) \leq 4 \exp(-\eta' N^{1-\beta} p_N) \leq 4 \exp(-\eta' \log^2 N).$$

This inequality governs  $\hat{\mu}_1(u) - \hat{\mu}(u)$  for each  $u$  in the ball  $\|u\| \leq N$ , but we can obtain a similar, uniform, inequality for the entire ball by checking  $\hat{\mu}_1 - \hat{\mu}$  at  $N^C$  points  $u$ ;  $C$  is a constant whose exact value is immaterial in the presence of the strong bound on  $P$ . This remark is valid for the remaining estimation of  $\hat{\mu}_1 - \hat{\mu}$ .

In the range  $N \leq \|u\| \leq N^2$  somewhat more care is required in the choice of  $t$ . The expected value of  $\exp t|\hat{\mu}_1 - \hat{\mu}|$  is bounded by  $4 \exp(Ct^2 p_N^{-1} \|u\|^{-1})$  if  $0 \leq t \leq \eta N p_N$ . For these numbers  $t$  we have

$$P(|\hat{\mu}_1(u) - \hat{\mu}(u)| \geq \varepsilon \|u\|^{-\beta/2}) \leq 4 \exp(Ct^2 p_N^{-1} \|u\|^{-1} - \varepsilon t \|u\|^{-\beta/2}).$$

The infimum, for unrestricted values  $t$ , obtained at  $t_0 = \varepsilon(2C)^{-1} p_N \|u\|^{1-\beta/2}$ , equals  $4 \exp(-\delta p_N \|u\|^{1-\beta}) \leq 4 \exp(-\delta \log^2 N)$ . This is of course a sufficient bound, and can be used if  $t_0 \leq \eta N p_N$ , that is,  $\|u\|^{1-\beta/2} \leq \eta_1 N$ . If we assume the opposite inequality,  $\|u\|^{1-\beta/2} \geq \eta_1 N$ , and choose  $t = \eta_2 N p_N$ , then the negative term exceeds twice the positive, and the bound becomes  $\exp(-\eta_3 N p_N \|u\|^{-\beta/2}) \leq \exp(-\eta_3 N p_N N^{-\beta}) = \exp(-\eta_3 \log^2 N)$ .

In the range  $\|u\| > N^2$ , we note that  $|b_k(u)| \leq C \|u\|^{-1/2}$  by the arguments used before in estimating variances, so the exponential bounds are valid for  $0 \leq t \leq C^{-1} \|u\|^{1/2} p_N$ . Now  $\|u\|^{1/2} p_N \|u\|^{-\beta/2} \geq \log^2 N$  for these values  $u$ , and the bounds of  $\hat{\mu} - \hat{\mu}_1$  can be extended to the ball  $\|u\| \leq N^C$ ,  $C = 4(1-\beta)^{-1}$ . But  $\sum \xi_k < 5N$  with probability near 1, so we have  $\sum |\xi_k - 1| \cdot |b_k(u)| < C \|u\|^{-1/2} 5N$  and this is  $O(\|u\|^{-\beta/2})$  when  $\|u\|^{1-\beta} > N^3$ , for example. Thus we obtain the required inequalities for  $\hat{\mu}(u) - \hat{\mu}_1(u)$  in four different regions, and the lemma is proved.

To prove the theorem in its entirety we begin with  $f \in C^2[-2, 2]$ ,  $f=0$  outside of  $(-1, 1)$ ,  $f>0$  on  $(-1, 1)$  and apply the modification of the lemma successively, with numbers  $\varepsilon_j = 3^{-j} \int f(x) dx$  so that the limit measure is positive. Its support  $S$  has finite Hausdorff measure for the function  $h(t) = t^\beta \log^2 t^{-1}$ , hence  $\dim S \leq \beta$ . The method can be improved to cover any measure function  $h(t)$  such that  $t^\beta \log t^{-1} = o(h)$  as  $t \rightarrow 0+$ . This approaches the theorem of Kahane [3, p. 13–15] in precision. In a certain sense, discussed briefly at the conclusion of [6], the set  $S$  is much more massive than some Salem sets of dimension  $\beta$ .

To obtain  $o(1)$  in place of  $O(1)$ , we observe that each measure in our construction belongs to  $C^2$ , so its Fourier transform is  $O(\|u\|^{-2})$ .

2. In the next theorem  $y$  is a function of class  $C^\infty(-\infty, \infty)$  and  $y'' > 0$  everywhere. Moreover  $\lambda$  is a probability measure on  $[0, 1]$  with a Lipschitz condition  $\lambda(a, a+h) \leq Ch^\alpha$  for a certain  $\alpha$  in  $(0, 1/2)$ . The measure  $\mu$  is now defined to be the image of  $\lambda$  by the mapping  $t \rightarrow (X(t), y \circ X(t))$  so that  $\mu$  is carried by the graph of  $y$  in the plane.  $X$  denotes Brownian motion.

**Theorem.** For almost all paths  $X$ ,  $|\hat{\mu}(u)| \leq C \|u\|^{\delta-\alpha}$  for  $u \in \mathbb{R}^2$ , and each  $\delta > 0$ .

For the proof we fix a function  $\Phi$  of class  $C^\infty$  and compact support and investigate, instead of  $\mu$ , the measure  $\Phi(X(t))\lambda(dt)$ , and its transform  $\mu_1$ . Our theorem will be a simple consequence of the next assertion:

(M) For each  $p=1, 2, 3, \dots$  the  $p$ -th moment of  $\hat{\mu}_1(u)$  admits a bound  $\|\hat{\mu}_1(u)\|_p^p \leq C \|u\|^{1-\alpha p}$ .

We observe that  $\Phi(x)=0$  outside a certain interval  $|x|\leq A$ , and on this interval we have  $|y'|\leq A_1$ . Now the transform  $\hat{\mu}_1(u)$ ,  $u=(u_1, u_2)$ , involves a mapping  $u_1x + u_2y(x)$ , whose derivative has absolute value  $\leq |u_1| - A_1|u_2|$ . As we shall soon observe, the estimation of  $\hat{\mu}_1(u)$  becomes much simpler if  $|u_1|\leq 2A_1|u_2|$ ; a similar easy case occurred in the proof completed above. In the opposite case,  $|u_2| > (2A_1)^{-1}|u_1|$ , we can write  $u_1x + u_2y(x) = \pm \|u\| g_u(x)$ , where  $g_u'' \geq c > 0$  and all derivatives of  $g_u$  are bounded on  $(-A, A)$  by constants independent of  $u$ . Until the conclusion of the demonstration of (M), we keep  $g = g_u$  and write  $v$  in place of  $\|u\|$ .

First of all we partition the interval  $0 < t < 1$  to isolate the small values of  $g'(X(t))$ . For this purpose we construct a function  $L$  of class  $C^\infty$ , vanishing outside  $(1, 4)$  so that  $0 \leq L \leq 1$  and  $\sum_{-\infty}^\infty L(2^k x) = 1$  if  $x > 0$ . A function  $L$  can be obtained from an ordinary partition of unity  $1 \equiv \sum h(y+k)$  by the substitution  $x = 2^y$ . We shall see that  $g'(X) = 0$  only on a set of  $\lambda$ -measure 0, so we have  $1 = \sum L(2^k g'(X)) + \sum L(-2^k g'(X))$  almost everywhere.

Let  $0 < \varepsilon < (8p)^{-1}$ ; we intend to neglect all the terms in the sum in which  $4^k \leq v^{1-\varepsilon}$ . The error introduced is no larger than

$$\lambda \{t : |g' X(t)| \leq C v^{\varepsilon/2} v^{-1/2}\} \quad \text{or}$$

$$\lambda \{t : |X(t) - x_0| \leq C' v^{\varepsilon/2} v^{-1/2}\} \quad \text{since}$$

$g'' \geq c > 0$ . The random variable  $h(r) = \sup \lambda \{t : |X(t) - x_0| \leq r\}$  has  $p$ -th moment  $\leq C r^{-1+2ap}$  [7], and therefore these terms can indeed be omitted from further calculations. We observe now that the number of terms remaining, in which  $L(2^k g'(X))\Phi(X)$  doesn't vanish identically, is at most  $C \log v$ . We complete the analysis for the integral containing  $L(2^k g')$ ; the method for  $L(-2^k g')$  is the same.

For each index  $k$  not already excluded we define  $r = r(k, v) = 4^k v^{\varepsilon-2}$  and divide the interval  $[0, 1]$  into adjacent subintervals of length  $r$ , denoting by  $I_j = I_j(k, v)$  the corresponding partial integrals. (Precisely,  $I_j$  is extended over the range  $jr \leq t \leq (j+1)r$ ). We shall use the theory of martingales to bound the moments of  $\sum I_{2j}$ ;  $\sum I_{2j+1}$  is handled in the same way. Now  $I_{2j}$  is measurable over the Borel field  $F_{2j+1} = F\{X(t); t \leq (2j+1)r\}$ . Thus we obtain a series of martingale differences by writing  $\sum I_{2j} - E(I_{2j}|F_{2j-1})$ , with the convention that  $F_{-1} = F_0$  is the trivial field. According to the Markov property  $E(I_{2j}|F_{2j-1}) = E(I_{2j}|X(2jr-r))$ , and we shall give a bound for this, uniform with respect to all values  $b = X(2jr-r)$ . We use the observation that when  $2jr \leq t \leq (2j+1)r$ ,  $X(t)$  has conditional distribution  $b + \lambda X(1)$ , with  $r \leq \lambda^2 \leq 2r$ . We are led to integrals of the form

$$\int \Phi(b + \lambda s) L(2^k g'(b + \lambda s)) \exp(-ivg(b + \lambda s)) e^{-s^2/2} ds / \sqrt{2\pi}.$$

The factor  $L(\cdot)$  vanishes outside the interval defined by the inequality  $1 < 2^k g'(b + \lambda s) < 4$ , and on this interval the derivative of  $vg(b + \lambda s)$  falls in the interval

$v2^{-k}\lambda < D < 4v2^{-k}\lambda$ . The successive derivatives of  $\Phi(b + \lambda s)$  remain bounded because  $\lambda \cong 2r^{1/2} = 2^k v^{-1} v^{e/2}$ , and we excluded all indices  $k$  for which  $4^k \cong v^{1-\varepsilon}$ . The  $m$ -th derivative of  $L(2^k g'(b + \lambda s))$  remains bounded for  $m=1, 2, 3, \dots$ , for the following reasons. For  $m=1$  we have the inequality  $2^k \lambda \cdot v 2^{-k} \lambda = v \lambda^2 \cong 2vr \cong 4^{k+1} v^{\varepsilon-1} < 4$ . From  $m \cong 2$  we observe that  $2^k \lambda^2 \cong 2^{k+1} r < 8^{k+1} v^{\varepsilon-2} < 8v^{3/2} v^{\varepsilon-2} < 8$ .

We intend to use  $z = 2^k \lambda^{-1} g(b + \lambda s)$  as a new variable; throughout the domain of integration  $1 < z' < 4$ . Moreover  $|z''| \cong C2^k \lambda \cong C2^{k+1} r^{1/2} \cong 2C4^k v^{\varepsilon/2} v^{-1} < 2Cv^{-\varepsilon/2}$ . Similar estimates apply to the higher derivatives of  $z$ ; the relations  $vg(b + \lambda s) = v\lambda 2^{-k} z$ ,  $v\lambda 2^{-k} > vr^{1/2} 2^{-k} = v \cdot 2^k v^{e/2} v^{-1} 2^{-k} = v^{e/2}$ , allow us to obtain the bound  $E(I_{2j}|F_{2j-1}) \cong C_B \mu(2jr, (2j+1)r)v^{-B}$  for any constant  $B$ .

We are now in sight of the moment inequality (M). We have just obtained a uniform bound on  $\sum E(I_{2j}|F_{2j-1})$ , and it remains to obtain bounds on the sum  $\sum I_{2j} - E(I_{2j}|F_{2j-1})$ . By an inequality of Burkholder [1, 2], in case  $p > 1$ , it will be sufficient to obtain bounds for  $\sum |I_{2j}|^2$  and  $\sum |E(I_{2j}|F_{2j-1})|^2$ . The second "square-function" is of course covered by the uniform estimates with a large exponent  $B$ . For the first sum we use the inequality  $(\sum |I|^2) \cong \max I \cdot \sum |I_j|$ . Recalling the dependence on  $k$  and  $v$ , we have  $\max I_j \cong Cr^\alpha = C4^{ak} v^{(\varepsilon-2)\alpha}$ . The sum  $\sum I_j$  doesn't exceed  $\int L(2^{-k} g' X(t)) \Phi(X(t)) d\lambda$ , and we saw before that this has  $p$ -th moment  $\cong C4^{-k\alpha p} 2^k$ . We find, then, that  $(\sum |I_j|^2)^{1/2}$  has  $p$ -th moment  $\cong C2^k v^{(\varepsilon-2)\alpha p} < Cv^{-2\alpha p} v$ . Summation with respect to  $k$  involves a factor  $\log^p v$ , but this can be absorbed by Schwarz' inequality. Thus (M) is proved.

Now (M) easily implies that  $|\hat{\mu}_1(u)| \cong C\|u\|^{\delta-\alpha}$  for characters  $u$  of the special form  $(\pm n_1^{1/2}, \pm n_2^{1/2})$ ,  $(n_1, n_2 = 0, 1, 2, \dots)$ . But  $\mu_1$  has compact support, and a device from Fourier analysis [3, p. 165] enables us to extend an inequality of this type to all of  $\mathbb{R}^2$ . By a suitable choice of a sequence  $\Phi_1, \Phi_2, \dots$ , we then obtain the theorem for  $\hat{\mu}(u)$ .

The inequality of Burkholder concerns martingales  $Y_n, 1 \cong n \cong N$ , their  $L^p$ -norm  $\|Y_N\|_p$ , and the  $L^p$ -norm of the square-function  $S$ , defined by  $S^2 = |Y_1|^2 + |Y_2 - Y_1|^2 + \dots + |Y_N - Y_{N-1}|^2$ . For  $1 < p < \infty$ ,  $\|S\|_p$  and  $\|Y_N\|_p$  are equivalent norms, with constants depending only on  $p$ .

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