# A MEASURE WHICH IS SINGULAR AND UNIFORMLY LOCALLY UNIFORM 

## by

David Freedman and Jim Pitman

Technical Report No. 180
(Revision of No. 163)

November 14, 1988

Department of Statistics
University of California
Berkeley, California

# A MEASURE WHICH IS SINGULAR AND UNIFORMLY LOCALLY UNIFORM 

by

David Freedman ${ }^{1}$<br>Jim Pitman ${ }^{2}$<br>Statistics Department<br>University of California<br>Berkeley, California 94720


#### Abstract

An example is given of a singular measure on [ 0,1 ] which is locally nearly uniform in the weak star topology. If this measure is used as a prior to estimate an unknown probability in coin tossing, the posterior is asymptotically normal.


## Keywords and phrases

Differentiation, Lebesgue points, Bayes estimates, Riesz product, singular measure, locally uniform measure, asymptotic normality of posterior distribution.

[^0]
## Introduction

Let $\mu$ be a probability on $[0,1]$. If $I$ is a subinterval of $[0,1]$, let $\mu_{I}$ be the probability on $[0,1]$ obtained from $\mu$ as follows: restrict $\mu$ to $I$; renormalize so the mass is 1 ; map $I$ affinely onto $[0,1]$, preserving the order. The image of the restricted and renormalized measure is $\mu_{I}$. To illustrate the notation, if $I=[a, b]$ then

$$
\mu_{I}[0, y]=\frac{\mu[a, a+y(b-a)]}{\mu[a, b]}, \text { for } y \in[0,1]
$$

Say that $\mu$ is locally uniform at $x$ when $x \in I$ and $|I| \rightarrow 0$ imply that $\mu_{I}$ converges to Lebesgue measure, in the weak star topology. Here $|I|$ is the length of $I$. In case the convergence of $\mu_{I}$ to Lebesgue measure as $|I| \rightarrow 0$ holds uniformly over all subintervals $I$, we call $\mu$ uniformly locally uniform. Using Kolmogorov's distance between probability distributions, this property can be expressed as follows:

$$
\begin{equation*}
\lim _{\delta \rightarrow 0} \sup _{|I| \leq \delta} \sup _{0 \leq y \leq 1}\left|\mu_{I}[0, y]-y\right|=0 \tag{1.1}
\end{equation*}
$$

A uniformly locally uniform measure is obviously continuous.
If $\mu$ is absolutely continuous with density $f$, then $f(x)>0$ and $x$ is a Lebesgue point of $f$ for $\mu$-almost all $x$; and $\mu$ is locally uniform at such $x$. See, for example, Dunford \& Schwartz (1958, pp217-8) or Saks (1937, Chap IV). It is natural to conjecture the converse: if $\mu$ is locally uniform at $\mu$-almost all $x$, it must be absolutely continuous. This is true, and easily proved, for local uniformity defined by convergence in variation norm. But for weak star convergence, this converse is false.
(1.2) Theorem. There exists a singular measure which is uniformly locally uniform.

Consider using a measure $\mu$ as a prior to estimate the unknown probability $p$ that a coin lands heads in a sequence of tosses. Let $\hat{p}$ be the fraction of heads among the first $n$ tosses. As shown by Laplace (1809), Bernstein (1917) and von Mises (1931), if the prior has a smooth density, the posterior is asymptotically normal. It may be conjectured that the converse holds: asymptotic normality of the posterior entails smoothness of the prior. But this too is false.
(1.3) Theorem. Suppose the prior $\mu$ is uniformly locally uniform. For every $\varepsilon>0$, the posterior distribution of $p \rightarrow \sqrt{\frac{n}{\hat{p}(1-\hat{p})}}(p-\hat{p})$ converges weak star to standard normal as $n \rightarrow \infty$, uniformly in $\hat{p} \in[\varepsilon, 1-\varepsilon]$.
If $0<p_{0}<1$, and the data are generated by tossing a $p_{0}$-coin, then $\hat{p} \rightarrow p_{0}$ almost surely, so the posterior is asymptoticlly normal almost surely. And the theorem applies to priors which may be singular, absolutely continuous, or mixed, provided
they are uniformly locally uniform. The argument also extends to give a condition on $\mu$ both necessary and sufficient for the conclusion of the theorem: for every $\varepsilon>0, \mu$ is uniformly locally uniform within $[\varepsilon, 1-\varepsilon$ ]. But we omit the details. There is an extensive literature on convergence of posterior distributions to normality in the stronger sense of total variation distance. See Le Cam (1986a and 1986b) for a survey.

The balance of this paper is organized as follows: Section 2 gives some reformulations of uniform local uniformity; Section 3 presents the construction for Theorem (1.2); Section 4 proves Theorem (1.3); Section 5 provides some background on Riesz products; finally, history and acknowledgements in Section 6.

## 2. A criterion for uniform local uniformity.

We start this section with an elementary estimate. This just shows that if $\mu$ puts approximately equal masses on equally spaced intervals, then $\mu$ will be close to uniform in Kolmogorov distance. The elementary proof is omitted.
(2.1) Lemma. Let $\mu$ be a probability on [0,1]. Let $0 \leq a<1,0<b<1$. Let $N$ be a positive integer, with $a+N b \leq 1<a+(N+1) b$, so the $N+1$ points

$$
a, a+b, a+2 b, \cdots, a+N b
$$

partition $[0,1]$ into $N$ subintervals of equal length $b$, and two shorter (and possibly degenerate) end intervals $[0, a)$ and $[a+N b, 1]$. Suppose that for every one of these $N+2$ intervals $I^{\prime}$, and every one of the $N$ intervals $I$ of length $b$,

$$
\frac{\mu\left(I^{\prime}\right)}{\mu(I)} \leq 1+\varepsilon
$$

Then

$$
\sup _{0 \leq y \leq 1}|\mu[0, y]-y| \leq \frac{4}{N}(1+\varepsilon)+\varepsilon
$$

We now give a criterion for $\mu$ to be uniformly locally uniform, in terms of the $\mu$ measure of adjacent binary intervals of order $k$, where $k$ is a positive integer, meaning intervals of the form

$$
I=\left[j / 2^{k},(j+1) / 2^{k}\right), \quad I^{\prime}=\left[j+1 / 2^{k},(j+2) / 2^{k}\right)
$$

for $j=0,1, \cdots, 2^{k}-2$ (except that $I^{\prime}$ should include 1 if that is its right end point).
(2.2) Proposition. A measure $\mu$ on [ 0,1 ] is uniformly locally uniform if and only if

$$
\begin{equation*}
\frac{\mu\left(I^{\prime}\right)}{\mu(I)} \rightarrow 1 \text { as } k \rightarrow \infty \tag{2.3}
\end{equation*}
$$

uniformly over all pairs I and $I^{\prime}$ of adjacent binary intervals of order $k$.
Proof. "Only if" is clear. For the converse, suppose (2.3) holds. Then (2.3) must also hold uniformly over all pairs $I$ and $I^{\prime}$ of binary intervals of order $k$ that are within distance $A 2^{-k}$ of each other, for each fixed $A>0$. If $J$ is any interval with with $2^{-i} \leq|J|<2^{-i+1}$, then $J$ contains at least $2^{j}-2$ consecutive binary intervals of order $k=i+j$, and is contained in a union of at most $2^{j+1}+2$ such intervals. Now a routine argument using Lemma (2.1) shows $\mu_{J}$ is close to uniform in Kolmogorov distance if $|J|$ is sufficiently small.

We conclude this section by stating two further conditions on a probability measure $\mu$ on [ 0,1 ], each implied by uniform local uniformity. These will be used in Section 4. Their proofs are elementary and omitted.
(2.4) Condition. Let $J$ and $J^{\prime}$ be two subintervals of an interval $I \subset[0,1]$. For each $\delta>0$,

$$
\frac{\mu\left(J^{\prime}\right) / \mu(J)}{\left|J^{\prime}\right| /|J|} \rightarrow 1 \text { as }|I| \rightarrow 0
$$

uniformly in $J, J^{\prime}, I$, provided

$$
\delta \leq\left|J^{\prime}\right| /|J| \leq 1 / \delta
$$

(2.5) Condition. Let $\mathbf{C}$ be a collection of uniformly bounded and uniformly equicontinuous functions on $[0,1]$. For a subinterval $I$ of $[0,1]$, let $A_{I}$ be the order preserving affine mapping of $I$ onto $[0,1]$. Then

$$
\frac{1}{\mu(I)} \int_{I} f \circ A_{I} d \mu \rightarrow \int_{0}^{1} f(x) d x
$$

as $|I| \rightarrow 0$, uniformly in $I$ and $f \in \mathbf{C}$.
Condition (2.4) is in fact equivalent to uniform local uniformity of $\mu$. So is (2.5) for sufficiently rich $\mathbf{C}$.

## 3. Proof of Theorem (1.2).

Let

$$
\begin{equation*}
f_{N}(x)=\prod_{1}^{N}\left[1+a_{n} \cos \left(2 \pi \lambda_{n} x\right)\right] \tag{3.1}
\end{equation*}
$$

where $0 \leq a_{n} \leq 1$, and the $\lambda_{n}$ are positive integers with integer ratios $\lambda_{n+1} / \lambda_{n} \geq 3$. Then $f_{N}$ is a probability density function. A probability measure $\mu$ on [0,1], called a Riesz product, can be defined as the weak star limit as $N \rightarrow \infty$ of the probability with density $f_{N}$. Informally,

$$
\begin{equation*}
\mu=\prod_{1}^{\infty}\left[1+a_{n} \cos \left(2 \pi \lambda_{n} x\right)\right] \tag{3.2}
\end{equation*}
$$

The existence of such Riesz product measures is shown by Fourier analysis. According to a result of Zygmund (1932; or 1959, p. 209), the Riesz product $\mu$ is either absolutely continuous or singular with respect to Lebesgue measure, according to the convergence or divergence of $\Sigma a_{n}^{2}$. In either case, $\mu$ is continuous. For the sake of completeness, in Section 5 we sketch arguments for the facts we use.

Because the assumptions imply that $\lambda_{1}$ divides $\lambda_{n}$ for every $n$,

$$
\begin{equation*}
\mu \text { has period } 1 / \lambda_{1} . \tag{3.3}
\end{equation*}
$$

That is, translating $\mu$ by $1 / \lambda_{1}(\bmod 1)$ leaves $\mu$ invariant. In particular,

$$
\begin{equation*}
\mu\left[j / \lambda_{1},(j+1) / \lambda_{1}\right)=1 / \lambda_{1} . \tag{3.4}
\end{equation*}
$$

(3.5) Lemma. Fix $N$. Suppose $I$ and $I^{\prime}$ are intervals of the form

$$
\left[j / \lambda_{n},(j+1) / \lambda_{n}\right), \quad\left[j^{\prime} / \lambda_{n},\left(j^{\prime}+1\right) / \lambda_{n}\right)
$$

with $n \leq N+1$ and $1 \leq j<j^{\prime}<\lambda_{n}$. If

$$
\begin{equation*}
b<\frac{f_{N}\left(x^{\prime}\right)}{f_{N}(x)}<c \quad \text { for all } x \text { in } I \text { and } x^{\prime} \text { in } I^{\prime} \tag{3.6}
\end{equation*}
$$

then

$$
\begin{equation*}
b<\frac{\mu\left(I^{\prime}\right)}{\mu(I)}<c \tag{3.7}
\end{equation*}
$$

Proof. Let

$$
\mu_{N}=\prod_{N+1}^{\infty}\left[1+a_{n} \cos \left(2 \pi \lambda_{n} x\right)\right] .
$$

Then $\mu$ has density $f_{N}$ with respect to $\mu_{N}$, so

$$
\mu(J)=\int_{J} f_{N} d \mu_{N} \quad \text { for } J=I \text { or } I^{\prime}
$$

Applying (3.4) to $\mu_{N}$ with $\lambda_{N+1}$ in place of $\lambda_{1}$ shows that $\mu_{N}(I)=\mu_{N}\left(I^{\prime}\right)$, because $\lambda_{n}$ divides $\lambda_{N+1}$. So (3.6) implies (3.7).

Fix now a sequence $a_{n}$ with $1 / 5 \geq a_{n} \rightarrow 0$ and $\Sigma a_{n}^{2}=\infty$, say $a_{n}=1 /(5 \sqrt{n})$. This makes the Riesz product $\mu$ continuous and singular. If the $\lambda_{n}$ increase rapidly enough, we can make $\mu$ uniformly locally uniform. The $\lambda_{n}$ will be of the form

$$
\begin{equation*}
\lambda_{n}=2^{k(n)} \tag{3.8}
\end{equation*}
$$

where $k(1)<k(2)<\cdots$ will be defined inductively. Let $\mu$ be the resulting Riesz product. The inductive definition of $k(n)$ will secure that (3.9) implies (3.10):
(3.9) $I$ and $I^{\prime}$ are two adjacent binary intervals of order $k$ with $k(n) \leq k \leq k(n+1)$,

$$
\begin{equation*}
1-4 a_{n} \leq \frac{\mu\left(I^{\prime}\right)}{\mu(I)} \leq 1+4 a_{n} \tag{3.10}
\end{equation*}
$$

This makes $\mu$ uniformly locally uniform by Proposition (2.2).

Inductive construction. Assume that $1=k(1)<\cdots<k(n)$ have been defined, hence also $\lambda_{1}, \cdots, \lambda_{n}$ via (3.8), and the partial Riesz product $f_{n}$ via (3.1). Use the strict positivity and uniform continuity of $f_{n}$ to choose $k(n+1) \geq k(n)+2$ so large that

$$
\begin{equation*}
1-a_{n+1}<\frac{f_{n}\left(x^{\prime}\right)}{f_{n}(x)}<1+a_{n+1} \text { provided }\left|x-x^{\prime}\right| \leq 2^{-k(n+1)+1} \tag{3.11}
\end{equation*}
$$

This completes the induction. Let $\mu$ be the Riesz product defined by (3.2). Take $n \geq 2$, so

$$
f_{n}(x)=f_{n-1}(x)\left[1+a_{n} \cos \left(2 \pi \lambda_{n} x\right)\right] .
$$

Thus

$$
\begin{equation*}
\left(1-a_{n}\right) f_{n-1}(x) \leq f_{n}(x) \leq\left(1+a_{n}\right) f_{n-1}(x) \tag{3.12}
\end{equation*}
$$

Assume $I$ and $I^{\prime}$ are as in (3.9). If $x \in I$ and $x \in I^{\prime}$, then

$$
\left|x-x^{\prime}\right| \leq 2^{-k+1} \leq 2^{-k(n)+1}
$$

We can use (3.11) with $n-1$ instead of $n$. This and (3.12) give the bounds

$$
\frac{\left(1-a_{n}\right)^{2}}{\left(1+a_{n}\right)}<\frac{f_{n}\left(x^{\prime}\right)}{f_{n}(x)}<\frac{\left(1+a_{n}\right)^{2}}{\left(1-a_{n}\right)}
$$

The same bounds for $\mu\left(I^{\prime}\right) / \mu(I)$ follow from Lemma (3.5). This yields (3.10), since
$a_{n} \leq 1 / 5$ by assumption. Conclusion: the measure $\mu$ is uniformly locally uniform.

## Remarks.

(i) Elementary estimates show that $f_{n}$ is bounded between $\exp ( \pm 3 \sqrt{n})$, and its logarithmic derivative is is bounded in absolute value by a constant times $\lambda_{n} \exp (6 \sqrt{n})$. Thus, (3.11) will be achieved provided

$$
2^{k(n)-k(n+1)} e^{6 \sqrt{n}} / a_{n+1} \rightarrow 0 \text { as } n \rightarrow \infty
$$

In particular, $k(n)$ of order $n^{\alpha}$ for $\alpha>3 / 2$ will do the job.
(ii) Given any sequence $\delta_{k}$ decreasing to zero as $k \rightarrow \infty$, replacing $2^{-k(n+1)+1}$ by $2^{-k(n+1)+1}+\delta_{k(n+1)}$ in (3.11) gives a singular $\mu$ such that $\mu\left(I^{\prime}\right) / \mu(I) \rightarrow 1$ as $k \rightarrow \infty$, uniformly over all pairs $I$ and $I^{\prime}$ of binary intervals of order $k$ within distance $\delta_{k}$ of each other. In other words, $\mu$ is nearly uniform over intervals which shrink to zero arbitrarily slowly.

## 4. Proof of Theorem (1.3).

In $n$ tosses of a p-coin, let $\hat{p}$ be the proportion of heads. The probability of getting any particular string of $j$ heads and $n-j$ tails is

$$
\begin{equation*}
L_{n}(p \mid \hat{p})=p^{j}(1-p)^{n-j}=\left[p^{\hat{p}}(1-p)^{1-\hat{p}}\right]^{n} \tag{4.1}
\end{equation*}
$$

Let $\mu$ be a prior distribution for $p$. For a Borel set $A \subset[0,1]$, let

$$
\begin{equation*}
\mu_{n}^{\#}(A \mid \hat{p})=\int_{A} L_{n}(p \mid \hat{p}) \mu(d p) \tag{4.2}
\end{equation*}
$$

The posterior distribution of $p$ is the probability $\tilde{\mu}_{n}(\cdot \mid \hat{p})$ on $[0,1]$ defined by

$$
\begin{equation*}
\tilde{\mu}_{n}(A \mid \hat{p})=\frac{\mu_{n}^{\#}(A \mid \hat{p})}{\mu_{n}^{\#}([0,1] \mid \hat{p})} \tag{4.3}
\end{equation*}
$$

Write

$$
\hat{\sigma}^{2}=\hat{p}(1-\hat{p}) / n .
$$

As is well known, the likelihood function $p \rightarrow \log L_{n}(p \mid \hat{p})$ is strictly concave; its maximum occurs at $p=\hat{p}$; and the function can be closely approximated by a Gaussian density with mean $\hat{p}$ and variance $\hat{\sigma}^{2}$, times a suitable scale factor. To be precise, we control the error in this approximation by the following lemma. Sharper estimates are given in Diaconis and Freedman (1988).
(4.4) Lemma.

$$
L_{n}(p \mid \hat{p})=L_{n}(\hat{p} \mid \hat{p}) \exp \left\{-\frac{1}{2} \frac{(p-\hat{p})^{2}}{\hat{\sigma}^{2}} F(p, \hat{p})\right\}
$$

where the factor $F(p, \hat{p})$ does not depend on $n$. For $\varepsilon \in(0,1 / 2)$,

$$
\begin{equation*}
F(p, \hat{p}) \rightarrow 1 \text { as }|p-\hat{p}| \rightarrow 0 \tag{4.5}
\end{equation*}
$$

uniformly over all $p, \hat{p} \in[\varepsilon, 1-\varepsilon]$. Finally

$$
\begin{equation*}
L_{n}(p \mid \hat{p}) \leq L_{n}(\hat{p} \mid \hat{p}) \exp \left\{-\frac{n}{2}(p-\hat{p})^{2}\right\} \tag{4.6}
\end{equation*}
$$

Proof. Let

$$
\begin{equation*}
H(p \mid \hat{p})=\frac{1}{n} \log L_{n}(p \mid \hat{p})=\hat{p} \log p+(1-\hat{p}) \log (1-p) . \tag{4.7}
\end{equation*}
$$

This does not depend on $n$. Now

$$
-\frac{d^{2}}{d p^{2}} H(p \mid \hat{p})=\frac{\hat{p}}{p^{2}}+\frac{1-\hat{p}}{(1-p)^{2}}=\dot{Q}(p, \hat{p}), \text { say }
$$

The function $p \rightarrow H(p \mid \hat{p})$ attains its maximum at $\hat{p}$. By Taylor's theorem, there is
an $r$ between $p$ and $\hat{p}$ such that

$$
H(p \mid \hat{p})=H(\hat{p} \mid \hat{p})-\frac{1}{2}(p-\hat{p})^{2} Q(r, \hat{p})
$$

Then

$$
\begin{aligned}
L_{n}(p \mid \hat{p}) & =\exp \{n H(p \mid \hat{p})\} \\
& =L_{n}(\hat{p} \mid \hat{p}) \exp \left\{-\frac{1}{2} \frac{(p-\hat{p})^{2}}{\hat{\sigma}^{2}} \hat{p}(1-\hat{p}) Q(r, \hat{p})\right\}
\end{aligned}
$$

Set $F(p, \hat{p})=\hat{p}(1-\hat{p}) Q(r, \hat{p})$. Relation (4.5) follows:

$$
|r-\hat{p}| \rightarrow 0 \text { as }|p-\hat{p}| \rightarrow 0 \text { and } Q(\hat{p}, \hat{p})=1 / \hat{p}(1-\hat{p}) .
$$

Finally, $Q(r, \hat{p}) \geq 1$; and this implies (4.6).
(4.8) Lemma. Assume that the prior $\mu$ is uniformly locally uniform. Fix $\varepsilon \in(0,1 / 2)$ and $0<\delta<K<\infty$. Then

$$
\mu_{n}^{\#}([\hat{p}-a \hat{\sigma}, \hat{p}+b \hat{\sigma}] \mid \hat{p}) \approx L_{n}(\hat{p} \mid \hat{p}) \mu(\hat{p}, \hat{p}+\hat{\sigma}) \sqrt{2 \pi}[\Phi(b)-\Phi(-a)]
$$

Here $\Phi$ is the standard normal distribution function. And the notation $\approx$ means that the ratio of the two sides tends to 1 as $n \rightarrow \infty$, uniformly over $\hat{p} \in[\varepsilon, 1-\varepsilon]$ and $0 \leq a, b, \leq K$ with $a+b \geq \delta>0$.
Proof. Let $I_{a b}=[\hat{p}-a \hat{\sigma}, \hat{p}+b \hat{\sigma}]$. Define a function $f_{a b}$ on $[0,1]$ by

$$
f_{a b}(x)=e^{-\frac{1}{2}[(a+b) x-a]^{2}}
$$

If $A$ is the affine mapping of $I_{a b}$ onto [ 0,1 ], then

$$
\begin{equation*}
\exp \left\{-\frac{1}{2} \frac{(p-\hat{p})^{2}}{\hat{\sigma}^{2}}\right\}=f_{a b}(A(p)) \text { for } p \in I_{a b} \tag{4.9}
\end{equation*}
$$

## Clearly,

(4.10) the $f_{a b}$ are bounded and uniformly equicontinuous as $a, b$ range over $[0, K]$.

We compute as follows:

$$
\begin{array}{rlr}
\mu_{n}^{\#}\left(I_{a b} \mid \hat{p}\right) & =\int_{I_{a b}} L_{n}(p \mid \hat{p}) \mu(d p) & \text { by definition of } \mu^{\#} \\
& \approx L_{n}(\hat{p} \mid \hat{p}) \int_{I_{a b}} \exp \left\{-\frac{1}{2} \frac{(p-\hat{p})^{2}}{\hat{\sigma}^{2}}\right\} \mu(d p) \quad \text { by (4.5) } \\
& \approx L_{n}(\hat{p} \mid \hat{p}) \mu\left(I_{a b}\right) \int_{0}^{1} f_{a b}(x) d x \quad \text { by (4.9), (4.10) and (2.5) }
\end{array}
$$

$$
=L_{n}(\hat{p} \mid \hat{p}) \frac{\mu\left(I_{a b}\right)}{a+b} \sqrt{2 \pi}[\Phi(b)-\Phi(-a)] \quad \text { by change of scale. }
$$

Use uniform local uniformity once more, as expressed in (2.4).
To complete the proof that the posterior distribution of $(p-\hat{p}) / \hat{\sigma}$ is asymptotically standard normal as $n \rightarrow \infty$, uniformly over $\hat{p} \in[\varepsilon, 1-\varepsilon]$, it only remains to establish the following lemma.
(4.11) Lemma. Suppose $\mu$ is uniformly locally uniform. Fix $\varepsilon, \delta>0$. There exists a $K=K(\varepsilon, \delta)$ such that

$$
\bar{\mu}_{n}([\hat{p}-K \hat{\sigma}, \hat{p}+K \hat{\sigma}] \mid \hat{p}) \geq 1-\delta \text { for all } \hat{p} \in[\varepsilon, 1-\varepsilon]
$$

for all $n$ greater than some $n(K, \varepsilon, \delta)$.
From (4.3), recall that $\bar{\mu}_{n}(\cdot \mid \hat{p})$ is the posterior distribution of $p$ given the first $n$ tosses.

Proof. We will argue that the posterior probability of ( $\hat{p}+K \hat{\sigma}, 1]$ is negligible. The argument for [ $0, \hat{p}-K \hat{\sigma}$ ). is symmetric. For $j=0,1, \cdots$, let

$$
I_{j}=(\hat{p}+j \hat{\sigma}, \hat{p}+(j+1) \hat{\sigma}] .
$$

Since $\mu$ is uniformly locally uniform, condition (2.4) implies that for all sufficiently large $n$,

$$
\mu\left(I_{j+1}\right) \leq 2 \mu\left(I_{j}\right) \quad \text { for all } j \text { and } \hat{p} .
$$

Thus

$$
\begin{equation*}
\mu\left(I_{j}\right) \leq 2^{j} \mu\left(I_{0}\right), \text { for } j \geq 1 \tag{4.12}
\end{equation*}
$$

Now for $\hat{p} \in[\varepsilon, 1-\varepsilon]$,

$$
\begin{align*}
\mu_{n}^{\#}\left(I_{j} \mid \hat{p}\right) & \leq \mu\left(I_{j}\right) L_{n}(\hat{p} \mid \hat{p}) e^{-\frac{1}{2} j^{2} \varepsilon(1-\varepsilon)}  \tag{4.6}\\
& \leq \mu\left(I_{0}\right) L_{n}(\hat{p} \mid \hat{p}) 2^{j} e^{-\frac{1}{2} j^{2} \varepsilon(1-\varepsilon)} \tag{4.12}
\end{align*}
$$

Summing this estimate over all $j \geq K$ gives

$$
\mu_{n}^{\#}((\hat{p}+K \hat{\sigma}, 1] \mid \hat{p}) \leq \mu\left(I_{0}\right) L_{n}(\hat{p} \mid \hat{p}) t(K, \varepsilon)
$$

where $t(K, \varepsilon)$ is the tail of a convergent series, so $t(K, \varepsilon) \rightarrow 0$ as $K \rightarrow \infty$ for every $\varepsilon$. On the other hand, using (4.8),

$$
\mu_{n}^{\#}([0,1] \mid \hat{p}) \geq \mu_{n}^{\#}\left(I_{0} \mid \hat{p}\right) \approx \mu\left(I_{0}\right) L_{n}(\hat{p} \mid \hat{p}) \sqrt{2 \pi}[\Phi(1)-\Phi(0)]
$$

Finally, use (4.3).

## 5. Existence and singularity of the Riesz product.

This section provides quick proofs from the literature of the two features of the Riesz product which are essential for the construction in Section 3.
Proof of existence of the Riesz product (3.1). (Adapted from Katznelson, 1976, p. 107). The condition $\lambda_{j+1} \geq 3 \lambda_{j}$ makes each integer $m$ have at most one representation of the form $\Sigma \xi_{j} \lambda_{j}$ where $\xi_{j}=-1,0,1$. If $m$ admits such a representation, and $|m| \leq \lambda_{N}$, then $\xi_{j}=0$ for all $j>N$. Now fix $m$ and $N$ with $|m| \leq \lambda_{N}$, and consider $n>N$. The $m$ th Fourier coefficient of the density $f_{n}$ in (3.1) is

$$
\begin{align*}
\int_{0}^{1} e^{-2 \pi i m x} f_{n}(x) d x & =\prod_{\xi_{j} \neq 0} \frac{1}{2} a_{j}, \text { if } m=\Sigma \xi_{j} \lambda_{j} \text { with } \xi_{j}=-1,0,1  \tag{5.1}\\
& =0, \text { if } m \text { does not admit this representation. }
\end{align*}
$$

Thus, as $n \rightarrow \infty$, every Fourier coefficient of $f_{n}$ is eventually constant. By a standard theorem, the probabilities with densities $f_{n}$ must therefore converge weak star to a probability $\mu$, whose nonzero Fourier coefficients are defined by the right side of (5.1).

Proof of singularity of the Riesz product (3.1) in case $\Sigma a_{n}^{2}=\infty$. (Adapted from Peyrière, 1973). We find from (5.1) that

$$
\begin{equation*}
\int_{0}^{1} e^{ \pm 2 \pi i \lambda_{n} x} \mu(d x)=\frac{1}{2} a_{n} \tag{5.2}
\end{equation*}
$$

and

$$
\begin{equation*}
\int_{0}^{1} e^{ \pm 2 \pi i \lambda_{n} x \pm 2 \pi i \lambda_{n^{\prime}} x} \mu(d x)=\frac{1}{2} a_{n} \frac{1}{2} a_{n^{\prime}} . \tag{5.3}
\end{equation*}
$$

In particular, the functions

$$
x \rightarrow e^{2 \pi i \lambda_{n} x}-\frac{1}{2} a_{n}
$$

are bounded and orthogonal in $L^{2}(\mu)$. Because $\Sigma a_{n}^{2}=\infty$, there must exist a sequence $c_{n}$ with

$$
\Sigma c_{n}^{2}<\infty, \quad c_{n} a_{n} \geq 0, \quad \Sigma c_{n} a_{n}=\infty
$$

(In our application, with $a_{n}=1 /(5 \sqrt{n}), c_{n}=1 /(\sqrt{n} \log n)$ will do.) But (5.2), (5.3) and $\Sigma c_{n}^{2}<\infty$ imply the series of functions

$$
\Sigma c_{n}\left(e^{2 \pi i \lambda_{n} x}-\frac{a_{n}}{2}\right)
$$

is an orthogonal convergent series in $L^{2}(\mu)$. On the other hand, the series

$$
\Sigma c_{n} e^{2 \pi i \lambda_{n} x}
$$

is orthogonal and convergent in $L^{2}$ (Lebesgue). By passing to subsequences, we can make the first series converge for $\mu$-almost all $x$, and the second for Lebesgue-almost all $x$. If $\mu$ were not singular, there would be an $x$ and a subsequence along which both series converged, hence also their difference would converge. But this is a contradiction, since their difference is $\frac{1}{2} \Sigma c_{n} a_{n}=\infty$.

The idea behind this and similar arguments of Brown and Moran $(1974,1975)$ is extended by Brown (1977), to give a general criterion for mutual singularity of probability measures based on comparison of sequences of square integrable random variables with low correlation.

## 6. History and Acknowledgements.

F. Riesz (1918) introduced the product (3.2) to exhibit various possible behaviours of Fourier coefficients at infinity. In particular, by taking $a_{j}=1$ for all $j$ and $\lambda_{j}=4^{j}$, he gave the first example of a continuous singular measure with Fourier coefficients not vanishing at infinity. See Graham and McGehee (1979, Chapter 7) for an extensive treatment of Riesz products in a more general setting.
In Freedman and Pitman (1988), we presented a continuous, singular probability measure $\mu$ on $[0,1]$ which is locally uniform at $\mu$-almost every point $x$ in [0, 1]. This $\mu$ was defined as the probability which makes the binary digits of $x$ independent, the $n^{\text {th }}$ digit being a one with probability $1 / 2+\varepsilon_{n}$ and a zero with probability $1 / 2-\varepsilon_{n}$, for a particular sequence $\varepsilon_{n}$ with $\varepsilon_{n} \rightarrow 0$ as $n \rightarrow \infty$ but $\Sigma \varepsilon_{n}^{2}=\infty$. Singularity follows from the criterion of Kakutani (1948). By working with such coin-tossing measures $\mu$ defined by various sequences $\varepsilon_{n}$, and introducing an element of smoothing into the construction, we subsequently became convinced that the exceptional set could be eliminated entirely, to yield a measure that was singular but uniformly locally uniform (SULU). But the construction was rather intricate. We are therefore very grateful to Russell Lyons, who suggested that such a measure might be created more easily as a Riesz product, and pointed us to the literature of these measures.

There is a close parallel between Riesz products and ordinary product measures of the coin-tossing kind mentioned above. For instance, it is easy to see that the coin-tossing measure for probabilities $1 / 2 \pm \varepsilon_{n}$ is the weak star limit as $N \rightarrow \infty$ of the probability with density

$$
\begin{equation*}
f_{N}(x)=\prod_{1}^{N}\left[1+2 \varepsilon_{n} r_{n}(x)\right], \tag{6.1}
\end{equation*}
$$

where $r_{n}(x)$ is the Rademacher function whose value is $\pm 1$ according to the $n$th binary digit of $x$. Compare with the definition of the Riesz product via (3.1): cosines instead of Rademacher functions, and $a_{n}$ instead of $2 \varepsilon_{n}$. Zygmund's dichotomy for Riesz products used in Section 3 thus corresponds to Kakutani's dichotomy for coin-tossing: the measures are either singular or absolutely continuous. The Rademacher functions are easier to deal with in some respects, since they are independent under Lebesgue measure, whereas the cosines are only orthogonal. But the smoothness of the cosines make the Riesz product easier to manipulate for present purposes.

## References

S. Bernstein (1917). Theory of Probability. (In Russian). Moscow.
G. Brown (1977). Singular infinitely divisible distributions whose characteristic functions vanish at infinity. Math. Proc. Camb. Phil. Soc. 82, 277-287.
G. Brown \& W. Moran (1974). On orthogonality of Riesz products. Proc. Camb. Phil. Soc. 76, 173-181.
G. Brown \& W. Moran (1975). Coin-tossing and powers of singular measures. Math. Proc. Camb. Phil. Soc. 77, 349-364.
P. Diaconis \& D. Freedman (1988). On the uniform consistency of Bayes estimates for multinomial probabilities. Technical report no. 137, Department of Statistics, University of California, Berkeley.
N. Dunford \& J.T. Schwartz (1958). Linear Operators. Part I: General Theory. Wiley Interscience, New York.
D. Freedman and J.W. Pitman (1988). A singular measure which is locally uniform. Technical report no. 163, Department of Statistics, University of California, Berkeley.
C.C. Graham \& O.C. McGehee (1979). Essays in Commutative Harmonic Analysis. Springer-Verlag, New York.
S. Kakutani (1948). On equivalence of infinite product measures. Ann. Math 49, 214-244.
Y. Katznelson (1968). An Introduction to Harmonic Analysis. Reprinted by Dover, New York, 1976.
P.S. Laplace (1809). Mémoire sur les intégrales définies et leur application aux probabilités, et spécialement à la recherche du milieu qu'il faut choisir entre les résultats des observations. Mémoires présentés à l'Académie des Sciences. Paris.
L. Le Cam (1986a). Asymptotic methods in statistical decision theory. Springer Verlag.
L. Le Cam (1986b). On the Bernstein-von Mises theorem. Technical report no. 57. Department of Statistics, University of California, Berkeley.
J. Peyrière (1973). Sur les produits de Riesz. C.R. Acad. Sci. Paris Sér A-B 276 1453-1455.
F. Riesz (1918). Uber die Fourierkoeffizienten einer stetigen Funktion von beschrankter Schwankung. Math. Zeitschr. 18, 312-315.
S. Saks (1937). Theory of the Integral. Reprinted in English by Dover, New York, 1964.
R. von Mises (1931). Wahrscheinlichkeitsrechnung. Springer Verlag, Berlin.
A. Zygmund (1932). On lacunary trigonmetric series. Trans. Amer. Math. Soc. 34, 435-446.
A. Zygmund (1959). Trigonometric Series, two volumes. Cambridge: The University Press.

# TECHNICAL REPORTS 

## Statistics Department

University of California, Berkeley

1. BREIMAN, L. and FREEDMAN, D. (Nov. 1981, revised Feb. 1982). How many variables should be entered in a regression equation? Jour. Amer. Statist. Assoc., March 1983, 78, No. 381, 131-136.
2. BRILLINGER, D. R. (Jan. 1982). Some contrasting examples of the time and frequency domain approaches to time series analysis. Time Series Methods in Hydrosciences, (A. H. El-Shaarawi and S. R. Esterby, eds.) Elsevier Scientific Publishing Co., Amsterdam, 1982, pp. 1-15.
3. DOKSUM, K. A. (Jan. 1982). On the performance of estimates in proportional hazard and log-linear models. Survival Analysis, (John Crowley and Richard A. Johnson, eds.) IMS Lecture Notes - Monograph Series, (Shanti S. Gupta, series ed.) 1982, 74-84.
4. BICKEL, P. J. and BREIMAN, L. (Feb. 1982). Sums of functions of nearest neighbor distances, moment bounds, limit theorems and a goodness of fit test. Ann. Prob., Feb. 1982, 11 . No. 1, 185-214.
5. BRILLINGER, D. R. and TUKEY, J. W. (March 1982). Spectrum estimation and system identification relying on a Fourier transform. The Collected Works of J. W. Tukey, vol. 2, Wadsworth, 1985, 1001-1141.
6. BERAN, R. (May 1982). Jackknife approximation to bootstrap estimates. Ann. Statist. March 1984, 12 No. 1, 101-118.
7. BICKEL, P. J. and FREEDMAN, D. A. (June 1982). Bootstrapping regression models with many parameters. Lehmann Festschrift, (P. J. Bickel, K. Doksum and J. L. Hodges, Jr., eds.) Wadsworth Press, Belmont, 1983, 28-48.
8. BICKEL, P. J. and COLLINS, J. (March 1982). Minimizing Fisher information over mixtures of distributions. Sankhya, 1983, 45, Series A. Pt. 1, 1-19.
9. BREIMAN, L. and FRIEDMAN, J. (July 1982). Estimating optimal transformations for multiple regression and correlation.
10. FREEDMAN, D. A. and PETERS, S. (July 1982, revised Aug. 1983). Bootstrapping a regression equation: some empirical results. JASA, 1984, 79, 97-106.
11. EATON, M. L. and FREEDMAN, D. A. (Sept. 1982). A remark on adjusting for covariates in multiple regression.
12. BICKEL, P. J. (April 1982). Minimax estimation of the mean of a mean of a normal distribution subject to doing well at a point. Recent Advances in Statistics, Academic Press, 1983.
13. FREEDMAN, D. A., ROTHENBERG, T. and SUTCH, R. (Oct. 1982). A review of a residential energy end use model.
14. BRILLINGER, D. and PREISLER, H. (Nov. 1982). Maximum likelihood estimation in a latent variable problem. Studies in Econometrics, Time Series, and Multivariate Statistics, (eds. S. Karlin, T. Amemiya, L. A. Goodman). Academic Press, New York, 1983, pp. 31-65.
15. BICKEL, P. J. (Nov. 1982). Robust regression based on infinitesimal neighborhoods. Ann. Statist., Dec. 1984, 12, 1349-1368.
16. DRAPER, D. C. (Feb. 1983). Rank-based robust analysis of linear models. I. Exposition and review.
17. DRAPER, D. C. (Feb 1983). Rank-based robust inference in regression models with several observations per cell.
18. FREEDMAN, D. A. and FIENBERG, S. (Feb. 1983, revised April 1983). Statistics and the scientific method, Comments on and reactions to Freedman, A rejoinder to Fienberg's comments. Springer New York 1985 Cohort Analysis in Social Research, (W. M. Mason and S. E. Fienberg, eds.).
19. FREEDMAN, D. A. and PETERS, S. C. (March 1983, revised Jan. 1984). Using the bootstrap to evaluate forecasting equations. J. of Forecasting. 1985, Vol. 4, 251-262.
20. FREEDMAN, D. A. and PETERS, S. C. (March 1983, revised Aug. 1983). Bootstrapping an econometric model: some empirical results. JBES, 1985, 2, 150-158.
21. FREEDMAN, D. A. (March 1983). Structural-equation models: a case study.
22. DAGGETT, R. S. and FREEDMAN, D. (April 1983, revised Sept. 1983). Econometrics and the law: a case study in the proof of antitrust damages. Proc. of the Berkeley Conference, in honor of Jerzy Neyman and Jack Kiefer. Vol I pp. 123-172. (L. Le Cam, R. Olshen eds.) Wadsworth, 1985.
23. DOKSUM, K. and YANDELL, B. (April 1983). Tests for exponentiality. Handbook of Statistics, (P. R. Krishnaiah and P. K. Sen, eds.) 4, 1984.
24. FREEDMAN, D. A. (May 1983). Comments on a paper by Markus.
25. FREEDMAN, D. (Oct. 1983, revised March 1984). On bootstrapping two-stage least-squares estimates in stationary linear models. Ann. Statist., 1984, 12, 827-842.
26. DOKSUM, K. A. (Dec. 1983). An extension of partial likelihood methods for proportional hazard models to general transformation models. Ann. Statist., 1987, 15, 325-345.
27. BICKEL, P. J., GOETZE, F. and VAN ZWET, W. R. (Jan. 1984). A simple analysis of third order efficiency of estimate Proc. of the Neyman-Kiefer Conference, (L. Le Cam, ed.) Wadsworth, 1985.
28. BICKEL, P. J. and FREEDMAN, D. A. Asymptotic normality and the bootstrap in stratified sampling. Ann. Statist. 12 470-482.
29. FREEDMAN, D. A. (Jan. 1984). The mean vs. the median: a case study in 4-R Act litigation. JBES. 1985 Vol 3 pp. 1-13.
30. STONE, C. J. (Feb. 1984). An asymptotically optimal window selection rule for kemel density estimates. Ann. Statist., Dec. 1984, 12, 1285-1297.
31. BREIMAN, L. (May 1984). Nail finders, edifices, and Oz.
32. STONE, C. J. (Oct. 1984). Additive regression and other nonparametric models. Ann. Statist., 1985, 13, 689-705.
33. STONE, C. J. (June 1984). An asymptotically optimal histogram selection rule. Proc. of the Berkeley Conf. in Honor of Jerzy Neyman and Jack Kiefer (L. Le Cam and R. A. Olshen, eds.), II, 513-520.
34. FREEDMAN, D. A. and NAVIDI, W. C. (Sept. 1984, revised Jan. 1985). Regression models for adjusting the 1980 Census. Statistical Science. Feb 1986, Vol. 1, No. 1, 3-39.
35. FREEDMAN, D. A. (Sept. 1984, revised Nov. 1984). De Finetti's theorem in continuous time.
36. DIACONIS, P. and FREEDMAN, D. (Oct. 1984). An elementary proof of Stirling's formula. Amer. Math Monthly. Feb 1986, Vol. 93, No. 2, 123-125.
37. LE CAM, L. (Nov. 1984). Sur l'approximation de familles de mesures par des familles Gaussiennes. Ann. Inst. Henri Poincaré, 1985, 21, 225-287.
38. DIACONIS, P. and FREEDMAN, D. A. (Nov. 1984). A note on weak star uniformities.
39. BREIMAN, L. and IHAKA, R. (Dec. 1984). Nonlinear discriminant analysis via SCALING and ACE.
40. STONE, C. J. (Jan. 1985). The dimensionality reduction principle for generalized additive models.
41. LE CAM, L. (Jan. 1985). On the normal approximation for sums of independent variables.
42. BICKEL, P. J. and YAHAV, J. A. (1985). On estimating the number of unseen species: how many executions were there?
43. BRILLINGER, D. R. (1985). The natural variability of vital raus and associated statistics. Biometrics, to appear.
44. BRILLINGER, D. R. (1985). Fourier inference: some methods for the analysis of array and nonGaussian series data. Water Resources Bulletin. 1985, 21, 743-756.
45. BREIMAN, L. and STONE, C. J. (1985). Broad spectrum estimates and confidence intervals for tail quantiles.
46. DABROWSKA, D. M. and DOKSUM, K. A. (1985, revised March 1987). Partial likelihood in transformation models with censored data.
47. HAYCOCK, K. A. and BRILLINGER, D. R. (November 1985). LIBDRB: A subroutine library for elementary time series analysis.
48. BRILLINGER, D. R. (October 1985). Fitting cosines: some procedures and some physical examples. Joshi Festschrifh 1986. D. Reidel.
49. BRILLINGER, D. R. (November 1985). What do seismology and neurophysiology have in common? - Statistics! Comptes Rendus Math. Rep. Acad. Sci. Canada. January, 1986.
50. COX, D. D. and O'SULLIVAN, F. (October 1985). Analysis of penalized likelihood-type estimators with application to generalized smoothing in Sobolev Spaces.
51. O'SULLIVAN, F. (November 1985). A practical perspective on ill-posed inverse problems: A review with some new developments. To appear in Journal of Statistical Science.
52. LE CAM, L. and YANG, G. L. (November 1985, revised March 1987). On the preservation of local asymptotic normality under information loss.
53. BLACKWELL, D. (November 1985). Approximate normality of large products.
54. FREEDMAN, D. A. (June 1987). As others see us: A case study in path analysis. Journal of Educational Statistics. 12, 101-128.
55. LE CAM, L. and YANG, G. L. (January 1986). Replaced by No. 68.
56. LE CAM, L. (February 1986). On the Bernstein - von Mises theorem.
57. O'SULLIVAN, F. (January 1986). Estimation of Densities and Hazards by the Method of Penalized likelihood.
58. ALDOUS, D. and DIACONIS, P. (February 1986). Strong Uniform Times and Finite Random Walks.
59. ALDOUS, D. (March 1986). On the Markov Chain simulation Method for Uniform Combinatorial Distributions and Simulated Annealing.
60. CHENG, C-S. (April 1986). An Optimization Problem with Applications to Optimal Design Theory.
61. CHENG, C-S., MAJUMDAR, D., STUFKEN, J. \& TURE, T. E. (May 1986, revised Jan 1987). Optimal step type design for comparing test treatments with a control.
62. CHENG, C-S. (May 1986, revised Jan. 1987). An Application of the Kiefer-Wolfowitz Equivalence Theorem.
63. O'SULLIVAN, F. (May 1986). Nonparametric Estimation in the Cox Proportional Hazards Model.
64. ALDOUS, D. (JUNE 1986). Finite-Time Implications of Relaxation Times for Stochastically Monotone Processes.
65. PITMAN, J. (JULY 1986, revised November 1986). Stationary Excursions.
66. DABROWSKA, D. and DOKSUM, K. (July 1986, revised November 1986). Estimates and confidence intervals for median and mean life in the proportional hazard model with censored data.
67. LE CAM, L. and YANG, G.L. (July 1986). Distinguished Statistics, Loss of information and a theorem of Robert B. Davies (Fourth edition).
68. STONE, C.J. (July 1986). Asymptotic properties of logspline density estimation.
69. BICKEL, P.J. and YAHAV, J.A. (July 1986). Richardson Extrapolation and the Bootstrap.
70. LEHMANN, E.L. (July 1986). Statistics - an overview.
71. STONE, C.J. (August 1986). A nomparametric framework for statistical modelling.
72. BIANE, PH. and YOR, M. (August 1986). A relation between Lévy's stochastic area formula, Legendre polynomial, and some continued fractions of Gauss.
73. LEHMANN, E.L. (August 1986, revised July 1987). Comparing Location Experiments.
74. O'SULLIVAN, F. (September 1986). Relative risk estimation.
75. O'SULLIVAN, F. (September 1986). Deconvolution of episodic hormone data.
76. PITMAN, J. \& YOR, M. (September 1987). Further asymptotic laws of planar Brownian motion.
77. FREEDMAN, D.A. \& ZEISEL, H. (November 1986). From mouse to man: The quantitative assessment of cancer risks To appear in Statistical Science.
78. BRILLINGER, D.R. (October 1986). Maximum likelihood analysis of spike trains of interacting nerve cells.
79. DABROWSKA, D.M. (November 1986). Nonparametric regression with censored survival time data.
80. DOKSUM, K.J. and LO, A.Y. (November 1986). Consistent and robust Bayes Procedures for Location based on Partial Information.
81. DABROWSKA, D.M., DOKSUM, K.A. and MIURA, R. (November 1986). Rank estimates in a class of semiparametric two-sample models.
82. BRILLINGER, D. (December 1986). Some statistical methods for random process data from seismology and neurophysiology.
83. DIACONIS, P. and FREEDMAN, D. (December 1986). A dozen de Finetti-style results in search of a theory. Ann. Inst. Henri Poincare, 1987, 23, 397-423.
84. DABROWSKA, D.M. (January 1987). Uniform consistency of nearest neighbour and kemel conditional Kaplan - Meier estimates.
85. FREEDMAN, D.A., NAVIDI, W. and PETERS, S.C. (February 1987). On the impact of variable selection in fiting regression equations.
86. ALDOUS, D. (February 1987, revised April 1987). Hashing with linear probing, under non-uniform probabilities.
87. DABROWSKA, D.M. and DOKSUM, K.A. (March 1987, revised January 1988). Estimating and testing in a two sample generalized odds rate model.
88. DABROWSKA, D.M. (March 1987). Rank tests for matched pair experiments with censored data.
89. DIACONIS, P and FREEDMAN, D.A. (April 1988). Conditional limit theorems for exponential families and finite versions of de Finetri's theorem. To appear in the Journal of Applied Probability.
90. DABROWSKA, D.M. (April 1987, revised September 1987). Kaplan-Meier estimate on the plane.

92a. ALDOUS, D. (April 1987). The Harmonic mean formula for probabilities of Unions: Applications to sparse random graphs.
93. DABROWSKA, D.M. (June 1987, revised Feb 1988). Nonparametric quantile regression with censored data.
94. DONOHO, D.L. \& STARK, P.B. (June 1987). Uncertainty principles and signal recovery.
95. CANCELLED
96. BRILLINGER, D.R. (June 1987). Some examples of the statistical analysis of seismological data. To appear in Proceedings, Centennial Anniversary Symposium, Seismographic Stations, University of California, Berkeley.
97. FREEDMAN, D.A. and NAVIDI, W. (June 1987). On the multi-stage model for carcinogenesis. To appear in Environmental Health Perspectives.
98. O'SULLIVAN, F. and WONG, T. (June 1987). Determining a function diffusion coefficient in the heat equation.
99. O'SULLIVAN, F. (June 1987). Constrained non-linear regularization with application to some system identification problems.
100. LE CAM, L. (July 1987, revised Nov 1987). On the standard asymptotic confidence ellipsoids of Wald.
101. DONOHO, D.L. and LIU, R.C. (July 1987). Pathologies of some minimum distance estimators. Annals of Statistics, June, 1988.
102. BRILLINGER, D.R., DOWNING, K.H. and GLAESER, R.M. (July 1987). Some statistical aspects of low-dose electron imaging of crystals.
103. LE CAM, L. (August 1987). Harald Cramér and sums of independent random variables.
104. DONOHO, A.W., DONOHO, D.L. and GASKO, M. (August 1987). Macspin: Dynamic graphics on a desktop computer. IEEE Computer Graphics and applications, June, 1988.
105. DONOHO, D.L. and LIU, R.C. (August 1987). On minimax estimation of linear functionals.
106. DABROWSKA, D.M. (August 1987). Kaplan-Meier estimate on the plane: weak convergence, LIL and the bootstrap.
107. CHENG, C-S. (Aug 1987, revised Oct 1988). Some orthogonal main-effect plans for asymmetrical factorials.
108. CHENG, C-S. and JACROUX, M. (August 1987). On the construction of trend-free run orders of two-level factorial designs.
109. KLASS, M.J. (August 1987). Maximizing $E \max _{1 \leq \mathrm{k} \leq \mathrm{n}} \mathrm{S}_{\mathbf{k}}^{+} / \mathrm{ES}_{\mathrm{n}}^{+}$: A prophet inequality for sums of I.I.D. mean zero variates.
110. DONOHO, D.L. and LIU, R.C. (August 1987). The "automatic" robustness of minimum distance functionals. Annals of Statistics, June, 1988.
111. BICKEL, P.J. and GHOSH, J.K. (August 1987, revised June 1988). A decomposition for the likelihood ratio statistic and the Bartlett correction - a Bayesian argument.
112. BURDZY, K., PITMAN, J.W. and YOR, M. (September 1987). Some asymptotic laws for crossings and excursions.
113. ADHIKARI, A. and PITMAN, J. (September 1987). The shortest planar arc of width 1.
114. RITOV, Y. (September 1987). Estimation in a linear regression model with censored data.
115. BICKEL, P.J. and RITOV, Y. (Sept. 1987, revised Aug 1988). Large sample theory of estimation in biased sampling regression models I.
116. RITOV, Y. and BICKEL, P.J. (Sept.1987, revised Aug. 1988). Achieving information bounds in non and semiparametric models.
117. RITOV, Y. (October 1987). On the convergence of a maximal correlation algorithm with alternating projections.
118. ALDOUS, D.J. (October 1987). Meeting times for independent Markov chains.
119. HESSE, C.H. (October 1987). An asymptotic expansion for the mean of the passage-time distribution of integrated Brownian Motion.
120. DONOHO, D. and LIU, R. (October 1987, revised March 1988). Geometrizing rates of convergence, II.
121. BRILLINGER, D.R. (October 1987). Estimating the chances of large earthquakes by radiocarbon dating and statistical modelling. To appear in Statistics a Guide to the Unknown.
122. ALDOUS, D., FLANNERY, B. and PALACIOS, J.L. (November 1987). Two applications of um processes: The fringe analysis of search trees and the simulation of quasi-stationary distributions of Markov chains.
123. DONOHO, D.L., MACGIBBON, B. and LIU, R.C. (Nov.1987, revised July 1988). Minimax risk for hyperrectangles.
124. ALDOUS, D. (November 1987). Stopping times and tightness II.
125. HESSE, C.H. (November 1987). The present state of a stochastic model for sedimentation.
126. DALANG, R.C. (December 1987, revised June 1988). Optimal stopping of two-parameter processes on nonstandard probability spaces.
127. Same as No. 133.
128. DONOHO, D. and GASKO, M. (December 1987). Multivariate generalizations of the median and trimmed mean II.
129. SMITH, D.L. (December 1987). Exponential bounds in Vapnik-Cervonenkis classes of index 1.
130. STONE, C.J. (Nov.1987, revised Sept. 1988). Uniform error bounds involving logspline models.
131. Same as No. 140
132. HESSE, C.H. (December 1987). A Bahadur - Type representation for empirical quantiles of a large class of stationary, possibly infinite - variance, linear processes
133. DONOHO, D.L. and GASKO, M. (December 1987). Multivariate generalizations of the median and trimmed mean, I.
134. DUBINS, L.E. and SCHWARZ, G. (December 1987). A sharp inequality for martingales and stopping-iimes.
135. FREEDMAN, D.A. and NAVIDI, W. (December 1987). On the risk of lung cancer for ex-smokers.
136. LE CAM, L. (January 1988). On some stochastic models of the effects of radiation on cell survival.
137. DIACONIS, P. and FREEDMAN, D.A. (April 1988). On the uniform consistency of Bayes estimates for multinomial probabilities.

137a. DONOHO, D.L. and LIU, R.C. (1987). Geometrizing rates of convergence, I.
138. DONOHO, D.L. and LIU, R.C. (January 1988). Geometrizing rates of convergence, III.
139. BERAN, R. (January 1988). Refining simultaneous confidence sets.
140. HESSE, C.H. (December 1987). Numerical and statistical aspects of neural networks.
141. BRILLINGER, D.R. (January 1988). Two reports on trend analysis: a) An Elementary Trend Analysis of Rio Negro Levels at Manaus, 1903-1985 b) Consistent Detection of a Monotonic Trend Superposed on a Stationary Time Series
142. DONOHO, D.L. (Jan. 1985, revised Jan. 1988). One-sided inference about functionals of a density.
143. DALANG, R.C. (February 1988). Randomization in the two-armed bandit problem.
144. DABROWSKA, D.M., DOKSUM, K.A. and SONG, J.K. (February 1988). Graphical comparisons of cumulative hazards for two populations.
145. ALDOUS, D.J. (February 1988). Lower bounds for covering times for reversible Markov Chains and random walks on graphs.
146. BICKEL, P.J. and RITOV, Y. (Feb.1988, revised August 1988). Estimating integrated squared density derivatives.
147. STARK, P.B. (March 1988). Strict bounds and applications.
148. DONOHO, D.L. and STARK, P.B. (March 1988). Rearrangements and smoothing.
149. NOLAN, D. (March 1988). Asymptotics for a multivariate location estimator.
150. SEILLIER, F. (March 1988). Sequential probability forecasts and the probability integral transform.
151. NOLAN, D. (March 1988). Limit theorems for a random convex set.
152. DIACONIS, P. and FREEDMAN, D.A. (April 1988). On a theorem of Kuchler and Laurizen.
153. DIACONIS, P. and FREEDMAN, D.A. (April 1988). On the problem of types.
154. DOKSUM, K.A. (May 1988). On the correspondence between models in binary regression analysis and survival analysis.
155. LEHMANN, E.L. (May 1988). Jerzy Neyman, 1894-1981.
156. ALDOUS, D.J. (May 1988). Stein's method in a two-dimensional coverage problem.
157. FAN, J. (June 1988). On the optimal rates of convergence for nomparametric deconvolution problem.
158. DABROWSKA, D. (June 1988). Signed-rank tests for censored matched pairs.
159. BERAN, R.J. and MILLAR, P.W. (June 1988). Multivariate symmetry models.
160. BERAN, R.J. and MILLAR, P.W. (June 1988). Tests of fit for logistic models.
161. BREIMAN, L. and PETERS, S. (June 1988). Comparing automatic bivariate smoothers (A public service enterprise).
162. FAN, J. (June 1988). Optimal global rates of convergence for nonparametric deconvolution problem.
163. DIACONIS, P. and FREEDMAN, D.A. (June 1988). A singular measure which is locally uniform.
164. BICKEL, P.J. and KRIEGER, A.M. (July 1988). Confidence bands for a distribution function using the bootstrap.
165. HESSE, C.H. (July 1988). New methods in the analysis of economic time series I.
166. FAN, JIANQING (July 1988). Nonparametric estimation of quadratic functionals in Gaussian white noise.
167. BREIMAN, L., STONE, C.J. and KOOPERBERG, C. (August 1988). Confidence bounds for extreme quantiles.
168. LE CAM, L. (August 1988). Maximum likelihood an introduction.
169. BREIMAN, L. (August 1988). Submodel selection and evaluation in regression-The conditional case and little bootstrap.
170. LE CAM, L. (September 1988). On the Prokhorov distance between the empirical process and the associated Gaussian bridge.
171. STONE, C.J. (September 1988). Large-sample inference for logspline models.
172. ADLER, R.J. and EPSTEIN, R. (September 1988). Intersection local times for infinite systems of planar brownian motions and for the brownian density process.
173. MILLAR, P.W. (October 1988). Optimal estimation in the non-parametric multipliçative intensity model.
174. YOR, M. (October 1988). Interwinings of Bessel processes.
175. ROJO, J. (October 1988). On the concept of tail-heaviness.
176. ABRAHAMS, D.M. and RIZZARDI, F. (September 1988). BLSS - The Berkeley interactive statistical system: An overview.
177. MILLAR, P.W. (October 1988). Gamma-funnels in the domain of a probability, with statistical implications.
178. DONOHO, D.L. and LIU, R.C. (October 1988). Hardest one-dimensional subfamilies.
179. DONOHO, D.L. and STARK, P.B. (October 1988). Recovery of sparse signals from data missing low frequencies.
180. FREEDMAN, D.A. and PITMAN, J.A. (Nov. 1988). A measure which is singular and uniformly locally uniform.
181. DOKSUM, K.A. and HOYLAND, ARNLJOT (Nov. 1988). A model for step-stress accelerated life testing experiments based on Wiener processes and the inverse Gaussian distribution.
182. DALANG, R.C., MORTON, A. and WILLINGER, W. (November 1988). Equivalent martingale measures and no-arbitrage in stochastic securities market models.

Copies of these Reports plus the most recent additions to the Technical Report series are available from the Statistics Department technical typist in room 379 Evans Hall or may be requested by mail from:

Department of Statistics
University of California
Berkeley, California 94720
Cost: \$1 per copy.


[^0]:    ${ }^{1}$ Research partially supported by NSF grant DMS 86-01634
    ${ }^{2}$ Research partially supported by NSF grant DMS 88-01808

