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FILTERING WITH SINGULAR CUMULATIVE SIGNALS

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Burgess Davis and Philip Protter

Mimeograph Series #81-8

Department of Statistics
Division of Mathematical Sciences
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April 1981

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Summary

Let $X = W + \alpha L$, where W is a Wiener process, L is the local time at 0 of either W or a Wiener process independent of W , and α is a constant. We show that if $|\alpha| \geq 16$ the minimal filtration of X contains the minimal filtration of W , that is, the "signal" αL can be completely recovered from observation of X . If α is small the problem is unsolved.

1. Introduction

A typical problem in filtering theory is that one observes a process $Y_t = W_t + \int_0^t h_s ds$, where W is a standard Wiener process representing the intergral of white noise, and one wants to estimate the "signal" h_t . In this paper we consider $Y^\alpha = W + \alpha L$ where L is local time at zero of W and α is a constant. We show in Theorem 3.2 that one has equality of the filtrations of Y^α and of W for $|\alpha| \geq 16$, and hence one need not make estimates at all, since αL can be completely recovered from observation of Y^α . We actually prove the stronger result that, for any $a < b$ and $|\alpha| \geq 16$,

$$(1.1) \quad \sigma(L_t - L_a, a \leq t \leq b) \subseteq \sigma(Y_t^\alpha - Y_a^\alpha, a \leq t \leq b).$$

With minor modifications our proof shows that (1.1) holds as well if L is local time at 0 of a Wiener process independent of W . By complicating the proof we can lower somewhat the bound 16, but we cannot get close to 1, and we conjecture that (1.1) does not hold for all positive α .

We thank J. Pitman for introducing us to these kinds of questions. He asked one of us ([3]) whether the filtrations of $W + L$ and W are identical, a question we have, evidently, not answered.

2. Theorems and Proofs

Throughout this paper we will take W to be a standard Wiener process defined on a complete probability space $(\Omega, \mathfrak{F}, P)$. $(\mathfrak{F}_t)_{t \geq 0}$ will denote the minimal completed filtration of W , and L will denote its local time at 0. We take L to be normalized so that

$$(2.1) \quad \lim_{\epsilon \rightarrow 0} \frac{1}{2\epsilon} \int_0^t 1_{(-\epsilon, \epsilon)}(W_s) ds = L_t$$

holds.

Define Y^α by

$$Y_t^\alpha = W_t + \alpha L_t$$

and let $(\mathcal{G}_t^\alpha)_{t \geq 0}$ be the minimal completed filtration for Y^α . Clearly $\mathcal{G}_t^\alpha \subseteq \mathfrak{F}_t$, $t \geq 0$.

(2.2) THEOREM. With the above notations $\mathcal{G}^\alpha = \mathfrak{F}$ for $|\alpha| \geq 16$.

First note we only need show $\mathfrak{F}_t \subseteq \mathcal{G}_t^\alpha$, $t \geq 0$. We consider only the case $\alpha = 16$; from the proof it is clear the same technique works for $|\alpha| > 16$. We let $\mathcal{G} = \mathcal{G}^\alpha$ and $Y = Y^\alpha$.

First a few facts needed later are collected. A standard estimate gives

$$(2.3) \quad P(W_1 > 5) < e^{-15},$$

and the following lemma is a standard large deviation result

(2.4) Lemma. If X_1, \dots, X_n are iid zero-one valued random variables with $P(X = 1) \leq e^{-15}$, then $P(\sum_{i=1}^n X_i > n/10) = o(2^{-n})$.

To prove this lemma we just take $t = 10$ in the inequality $P(\sum X_i > n/10) \leq E(e^{t \sum X_i}) / e^{t(n/10)}$.

Next fix $t > 0$ and let H be an open interval in $[0, t]$. We will construct a zero-one valued r.v. T , \mathcal{G}_t -measurable, such that $\{T = 0\} = \{\omega: B_s = 0 \text{ in } H\}$ a.s. This will allow us to recover L in a \mathcal{G} -optional way.

Divide H into m equal disjoint subintervals $(\Delta_p)_{1 \leq p \leq m}$ so that

$$H = \bigcup_{p=1, m} \Delta_p.$$

Fix $p < m$ and write $\Delta = \Delta_p = (x, x + \gamma)$. Divide Δ into 2^n equal disjoint subintervals $(J_q)_{1 \leq q \leq 2^n}$ so that:

$$\Delta_p = \Delta = \bigcup_{q=1, 2^n} J_q.$$

The length of $J_q = |J_q| = \gamma/2^n$, and letting $\epsilon = \gamma/2^n$ we form the following intervals for each $J_q = (a, b)$:

$$I_1^q = (a, a + \epsilon) = (a_1^q, b_1^q) = (a_1, b_1)$$

⋮

$$I_n^q = (a + 2^{n-1}\epsilon, a + 2^n\epsilon) = (a_n^q, b_n^q) = (a_n, b_n).$$

Recalling $\Delta = (x, x + \gamma)$, note that each $J_i \subseteq (x, x + 2\gamma) \subseteq H$, for Δ_p with $p < m$. Finally, we define r.v.'s $Z(J_q, n)$ by:

$$(2.5) \quad Z(J_q, n) = \begin{cases} 1 & \text{if } \sum_{i=1, n} 1_{\{Y_{b_i} - Y_{a_i} > 5 \sqrt{b_i - a_i}\}} > n/10 \\ 0 & \text{otherwise.} \end{cases}$$

(2.6) LEMMA. $P(Z(J_q, n) = 1 \cap \{\omega: W_s = 0 \text{ in } (a_1^q, b_n^q)\}) = o(2^{-n})$.

Proof of Lemma 2.6: Let $\Lambda_q = \{\omega: W_s = 0 \text{ in } (a_1^q, b_n^q)\}$. Note that L is constant on Λ_q ; hence

$$\begin{aligned}
 P(Z(J_q, n) = 1 \cap \Lambda_q) &\leq P(\sum_{i=1, n}^1 (W_{b_i} - W_{a_i}) > 5 \sqrt{b_i - a_i}) \sim \frac{n}{10} \\
 &= P(\sum_{i=1, n} X_i > n/10)
 \end{aligned}$$

and an application of Lemma 2.4 completes the proof.

(2.7) LEMMA. $P(Y_{s+1/2} - Y_s > 5 \frac{1}{\sqrt{2}}) \geq .11$, for any s , $0 \leq s < 1/2$.

Proof of Lemma 2.7: Suppose we can show $P(L_{s+1/2} - L_s > .37) \geq .112$ for any s , $0 \leq s < 1/2$. Then

$$\begin{aligned}
 P(Y_{s+1/2} - Y_s > 5 \frac{1}{\sqrt{2}}) \\
 &\geq P(16L_{s+1/2} - 16L_s > 5.8) - P(W_{1/2} < -2.2) \\
 &\geq .112 - .002 = .11.
 \end{aligned}$$

Let $R = \inf\{t > s : W_t = 0\}$. Then

$$\begin{aligned}
 P(L_{s+1/2} - L_s > .37) &\geq P(L_{R+1/6} - L_R > .37 | R \leq s + 1/3) \cdot P(R \leq s + 1/3) \\
 &\geq .35P(R \leq s + 1/3),
 \end{aligned}$$

since $P(L_{R+1/6} - L_R > .37 | R \leq s + 1/3) \geq P(L_{1/6} > .37) = 2P(W_{1/6} > .37)$. It remains to show $P(R \leq s + 1/3) \geq .32$. Then $P(R \leq s + 1/3) =$

$$\begin{aligned}
 &E\{P(B \text{ hits } 0 \text{ in } (s, s + 1/3) | B_s)\} \\
 &= E\{2(1 - \Phi_{1/3}(|B_s|))\} \\
 &= 4 \int_0^\infty (1 - \Phi_{1/3}(y)) \varphi_s(y) dy \quad [0 \leq s < 1/2] \\
 &\geq 4 \int_0^\infty (1 - \Phi_{1/3}(y)) \varphi_{1/2}(y) dy \\
 &\geq .32,
 \end{aligned}$$

where Φ and φ are the usual Gaussian cdf and pdf.

□

Set $\Gamma_q = \{\omega: B_s = 0 \text{ in } \cup I_n^q\}$. The last lemma immediately implies

$$(2.8) \quad P(Y_{b_i} - Y_{a_i} > 5 \sqrt{b_i - a_i} | \Gamma_q) > .11, \text{ for } 1 \leq i \leq n$$

Using (2.8) observe that:

$$\begin{aligned} (2.11) \quad n &\leq E\{\sum_{i=1, n}^1 (Y_{b_i} - Y_{a_i} > 5 \sqrt{b_i - a_i}) | \Gamma_q\} \\ &\leq nP(\sum_{i=1, n}^1 (Y_{b_i} - Y_{a_i} > 5 \sqrt{b_i - a_i}) > n/10) + n/10; \end{aligned}$$

hence

$$(2.9) \quad P(Z(J_{q,n}) = 1 | \Gamma_q) > .01.$$

Recalling that p is fixed, define

$$(2.10) \quad R_{p,n} = \sup_{1 \leq q \leq 2^n} Z(J_{q,n}).$$

Let

$$(2.11) \quad A_p = \{\omega: W_s = 0 \text{ in } \Delta_p\}, \text{ and}$$

$$A_{p,q_0} = \{J_{q_0,n} \text{ is the first } J_q \text{ such that} \\ \omega \text{ with } W_s = 0 \text{ in } J_{q,n}\}.$$

Then

$$\begin{aligned} (2.12) \quad P(R_{p,n} = 1 | A_p) &\geq \sum_{q=1}^n P(Z(J_{q,n}) = 1 | A_{p,q}) P(A_{p,q} | A_p) \\ &> (.01) \sum_{q=1}^{2^n} P(A_{p,q} | A_p) = .01 \end{aligned}$$

where we have used (2.9). Next, letting

$$(2.13) \quad D_p = \{\omega: W_s = 0 \text{ in } \Delta_p \cup \Delta_{p+1}\}$$

we have that

$$\begin{aligned} P(R_{p,n} = 1 | D_p) &\leq \sum_{q=1, 2^n} P(Z(J_{q,n}) = 1 | D_p) \\ &= 2^n o(2^{-n}) = o(1) \end{aligned}$$

by Lemma 2.6, hence

$$c_n = P(R_{p,n} = 1 \cap D_p) = o(1).$$

Choose a subsequence $\{n_i\}$ such that $\sum_{i=1, \infty} c_{n_i} < \infty$, for c_n as above.

Define

$$(2.14) \quad R_p = \limsup_{i \rightarrow \infty} R_{p, n_i},$$

with R_p zero-one valued. By the definition of R_p and Borel-Cantelli we have

$$(2.15) \quad P(R_p = 1 \cap D_p) = 0$$

while

$$(2.16) \quad P(R_p = 1 | A_p) \geq .01$$

from (2.12) and Fatou's lemma. Note that R_p depends only on the values of Y in $\Delta_p \cup \Delta_{p+1}$. We now define

$$(2.17) \quad T_m = \sup_{1 \leq p \leq m-1} R_p.$$

Let

$$(2.18) \quad \begin{cases} A = \{\exists s: \text{ such that } W_s = 0 \text{ in } H\} \\ D = \{\exists s: \text{ such that } W_s = 0 \text{ in } H\}. \end{cases}$$

Then it follows easily from (2.15) and (2.16) that

$$(2.19) \quad \begin{cases} P(T_m = 1 \cap D) = 0 \\ P(T_m = 1 | A) \geq .01. \end{cases}$$

Next we set

$$(2.20) \quad T'_H = \limsup_{m \rightarrow \infty} T_m.$$

Using (2.19) note that Borel-Cantelli and Fatou's lemma respectively yield

$$P(\{T'_H = 1\} \cap D) = 0$$

$$P(T'_H = 1 | A) \geq .01.$$

We note that $\{T'_H = 1\}$ is measurable $\sigma(Y_t - Y_a, a \leq t \leq b + \epsilon)$ for any ϵ .

Secondly, the proof just given immediately generalizes to show that

$$(2.21) \quad P(T_H^1 = 1 | A \cap B) \geq .01,$$

where B is any set in $\sigma(Y_t, t \leq a)$. Now let (c_n, d_n) , $n \geq 1$, be an ordering of all the open intervals contained in (a, b) which have rational endpoints.

Define $T_H = T = \sup_n T_{(c_n, d_n)}^1$. Then (2.21) immediately implies

$$(2.22) \quad P(T_H = 1 | A) = 1,$$

since if W_t hits 0 in H it is possible to find with probability 1, an arbitrarily large finite disjoint collection of intervals (c_k, d_k) such that W_t hits 0 in each of the intervals.

At the beginning of the proof (following Lemma 2.4) we fixed an open interval H in $[0, t]$. For each such H with rational endpoints, $\{T_H = 0\} = \{\forall s: W_s = 0 \text{ in } H\}$ a.s., and $\{T_H = 0\} \in \mathcal{G}_t$. Thus the complement of the zero set of W is a \mathcal{G}^α -optional set. Then $H_s = 1_Z(s)$, the indicator of the zero set of W , is \mathcal{G}^α -predictable and $W_t = \int_0^t \frac{1}{\alpha} H_s dY_s$. Alternatively, a theorem due to Lévy allows us to express the local time L as a limit of these random intervals (cf [2, p. 730]). Thus L is a \mathcal{G}^α -optional process and this completes the proof of Theorem 3.2.

□

(2.23) COMMENT. Filtering theory is usually concerned with cumulative signal processes that have absolutely continuous paths. If $Y_t = W_t + \int_0^t h(s, W_s) ds$ for bounded, Borel h , then using Girsanov's theorem one can find a new probability law Q that makes Y into a Brownian motion. W is then a solution of the differential equation

$$dW_s = dY_s - h(s, W_s) ds$$

and W is a strong solution by Zvonkin's theorem [5]. Thus in this case as well the filtrations of Y and of W are the same and one need not make estimates

of $h_s = h(s, W_s)$. Indeed, we do not know of any additive functionals A of W such that if $Y = W + A$ then the filtrations of Y and W are different; we conjecture, however, that if $A = \epsilon L$ for small enough ϵ then the filtrations are in fact different.

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