## Stochastic Filtering（SS2016）Exercise Sheet 8

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## 8．1．Kushner－Stratonovich and Zakai equation

Let $X$ solve the martingale problem associated to $A: \mathscr{D}(A) \subseteq B\left(\mathbb{R}^{d}\right) \rightarrow B\left(\mathbb{R}^{d}\right)$ and let $d Y_{t}=h\left(X_{t}\right) d t+d W_{t}$ ．Recall that for any bounded measurable function $f: \mathbb{X} \rightarrow \mathbb{R}, \rho(f)$ is defined as the $(\mathbb{F}(Y), \tilde{\mathbb{P}})$－optional projection of $\Lambda f(X)$ ，where

$$
\begin{array}{ll}
\left.\mathbb{P}\right|_{\mathscr{F}_{t}}=\left.\Lambda_{t} \tilde{\mathbb{P}}\right|_{\mathscr{F}_{t}}, & \Lambda=\mathscr{E}(h(X) \bullet Y), \\
\left.\tilde{\mathbb{P}}\right|_{\mathscr{F}_{t}}=\left.Z_{t} \mathbb{P}\right|_{\mathscr{F}_{t}}, & Z=\mathscr{E}(-h(X) \bullet W) .
\end{array}
$$

a）Assume that $h$ is bounded．Show that $\rho(1)=\mathscr{E}(\pi(h) \bullet Y)$ ．
Hint：Show that for any bounded stopping time $T$

$$
\tilde{\mathbb{E}}\left[\mathbb{1}_{T<\infty} \Lambda_{T}\right]=\tilde{\mathbb{E}}\left[\mathbb{1}_{T<\infty} \int_{0}^{T} \pi_{s}(h) \rho_{s}(1) d Y_{s}\right]
$$

by applying the martingale representation theorem to $\mathbb{1}_{T<\infty}$ ．
b）Deduce the Zakai equation from the Kushner－Stratonovich equation and a）．

## 8．2．Change of measure approach for jump processes

Let $X$ solve the martingale problem associated to $A: \mathscr{D}(A) \subseteq B\left(\mathbb{R}^{d}\right) \rightarrow B\left(\mathbb{R}^{d}\right)$ and let $Y$ be a Poisson process with rate $\lambda\left(X_{-}\right)$，i．e．，$Y_{t}=N_{\int_{0}^{t} \lambda\left(X_{s-}\right) d s}$ ，where $N$ is a standard Poisson process independent of $X$ and $\lambda: \mathbb{X} \rightarrow(0, \infty)$ is a measurable function．Assume that $\lambda$ and $\lambda^{-1}$ are bounded．

Define a change of measure from $\mathbb{P}$ to $\tilde{\mathbb{P}}$ via

$$
\begin{array}{ll}
\left.\mathbb{P}\right|_{\mathscr{F}_{t}}=\left.\Lambda_{t} \tilde{\mathbb{P}}\right|_{\mathscr{F}_{t}}, & d \Lambda_{t}=\Lambda_{t-}\left(\lambda\left(X_{t-}\right)-1\right)\left(d Y_{t}-d t\right), \\
\left.\tilde{\mathbb{P}}\right|_{\mathscr{F}_{t}}=\left.Z_{t} \mathbb{P}\right|_{\mathscr{F}_{t}}, & d Z_{t}=Z_{t-}\left(\lambda^{-1}\left(X_{t-}\right)-1\right)\left(d Y_{t}-\lambda\left(X_{t-}\right) d t\right) .
\end{array}
$$

It can be shown that the law of $(X, Y)$ under $\tilde{\mathbb{P}}$ equals the law of $(X, N)$ under $\mathbb{P}$.
a) Show that $\left[M^{f}, Y\right]=0$, where

$$
M_{t}^{f}=f\left(X_{t}\right)-f\left(X_{0}\right)-\int_{0}^{t} A f\left(X_{s-}\right) d s
$$

Sketch of proof: Let $\left(T_{n}\right)_{n \in \mathbb{N}}$ be a sequence of stopping times exhausting the jumps of $M^{f}$. Under the measure $\tilde{\mathbb{P}}$, the process $Y$ is a standard Poisson process independent of $X$. Therefore, it has no fixed times of discontinuity, i.e., $\tilde{\mathbb{P}}\left[Y_{t} \neq 0\right]=0$ holds for each deterministic time $t \in \mathbb{R}$. Together with the independence of $T_{n}$ and $Y$ this implies $\tilde{\mathbb{P}}\left[\Delta Y_{T_{n}} \neq 0\right]=0$ for each $n \in \mathbb{N}$. Thus, $M^{f}$ and $Y$ have no common jumps and $\left[M^{f}, Y\right]=0$.
b) Derive the Zakai equation, i.e.,

$$
d \rho_{t}(f)=\rho_{t}(A f) d t+\left(\rho_{t-}(\lambda f)-\rho_{t-}(f)\right)\left(d Y_{t}-d t\right)
$$

Hint: You can use exactly the same steps as in the lecture, where Zakai's equation was derived for observations with additive Gaussian noise.

### 8.3. Singular filtering and stochastic volatility

This example shows why one typically assumes that the volatility of the observational noise does not depend on the signal process $X$.

We work on a filtered probability space $(\Omega, \mathscr{F}, \mathbb{F}, \mathbb{P})$ satisfying the usual conditions. The signal is a càdlàg, $\mathbb{F}$-adapted, $\mathbb{X}$-valued process $X$. The observation process $Y$ satisfies $Y_{0}=0$ and

$$
d Y_{t}=h\left(X_{t}\right) d t+\sigma\left(X_{t-}\right) d W_{t},
$$

where $W$ is a standard $\mathbb{F}$-Wiener process, $\sigma: \mathbb{X} \rightarrow[0, \infty)$ and $h: \mathbb{X} \rightarrow \mathbb{R}$ are both continuous.
a) Suppose $\mathbb{X}=[0, \infty)$ and $\sigma$ is bijective. Derive an expression for the filter $\pi$.

Hint: Show that $X$ is $\mathbb{F}(Y)$-adapted.
b) Let $T>0$ be fixed and consider an increasing sequence $\left(A_{n}\right)_{n \in \mathbb{N}}$ of finite subsets of $[0, T]$ such that $\bigcup_{n \in \mathbb{N}} A_{n}$ is dense in $[0, T]$. Suppose that the process $Y$ from a) is observable only at the time-points $t \in A_{n}$. Let $\mathscr{G}_{n}$ be the (augmented) $\sigma$-algebra generated by $\left\{Y_{t}: t \in A_{n}\right\}$. Show that for any bounded $f$, we have

$$
\lim _{n \rightarrow \infty} \mathbb{E}\left[f\left(X_{T}\right) \mid \mathscr{G}_{n}\right]=\pi_{T}(f) \quad \text { a.s. }
$$

Hint: Apply Doob's martingale convergence theorem.
c) Compare the results of this exercise to Exercise 7.4. Give a mathematically precise interpretation of the following sentence: "High-frequency data allows one to estimate volatilities, but not drifts."

### 8.4. Singular filtering of a two-dimensional process

This is an example of a singular filtering problem with an explicit solution. Consider the setting of Exercise 8.3 with $\mathbb{X}=\mathbb{R}^{2}, h(x)=0, \sigma(x)=\|x\|$ for $x \in \mathbb{R}^{2}$, and $X$ is a 2dimensional standard Wiener process independent of $W$. Calculate the filter of $X$ given $Y$.

Hint: Show that $\pi_{t}$ is a spherical distribution on $\mathbb{R}^{2}$, i.e., $\pi_{t}(f)=\pi_{t}(f \circ U)$, where $f$ : $\mathbb{R}^{2} \rightarrow \mathbb{R}$ is bounded and $U$ denotes multiplication by an orthogonal $2 \times 2$ matrix. Use that any spherical distribution on $\mathbb{R}^{2}$ can be represented as the law of $R S$, where $R$ is a random variable with values in $[0, \infty)$ and $S$ is uniformly distributed on the unit sphere in $\mathbb{R}^{2}$, independent of $R$.

