

# Introduction to stochastic differential equations – 2

## Stochastic exponentials, Girsanov's theorems

Alexander Veretennikov<sup>1</sup>  
Spring 2020

April 15, 2020

# Abstract

First example of weak solutions; no Lipschitz conditions

A stochastic differential equation in  $\mathbb{R}^d$  is considered

$$dX_t = b(t, X_t)dt + dW_t, \quad t \geq 0, \quad X_0 = x_0, \quad (1)$$

Or, equivalently in the integral form,

$$X_t = x_0 + \int_0^t b(s, X_s)ds + W_t. \quad (2)$$

Here  $(W_t, \mathcal{F}_t)$  is a standard  $d$ -dimensional Wiener process,  $b$  and  $\sigma$  are vector and matrix Borel functions of corresponding dimensions  $d$  and  $d \times d$ . The initial value  $x_0$  may be non-random, or random but  $\mathcal{F}_0$ -measurable. Yet, the function  $b$  is only Borel measurable and bounded. Is there a solution?

# Stochastic exponentials

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Let  $b_t$  be an adapted bounded stochastic vector-valued  $d$ -dimensional process. Denote

$$\rho_t = \rho_t[b] := \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right),$$

where  $b^2 := (b, b)$ , that is, a scalar product. By Ito's formula,

$$d\rho_t = b_t \rho_t dW_t \quad (\text{here } b dW_t \text{ is also a scalar product}).$$

In other words,  $\rho_t$  is a solution of an SDE *with a random diffusion coefficient*

$$dX_t = b_t X_t dW_t, \quad X_0 = 1.$$

In the integral form we have,

$$\rho_t = 1 + \int_0^t b_s \rho_s dW_s.$$

$$\rho_t = \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right)$$

$$\rho_t = 1 + \int_0^t b_s \rho_s dW_s$$

The integral form gives us a hope that possibly  $\rho_t$  may be a martingale, as usual for stochastic integral. If this is true, we would have, in particular,  $E\rho_t = 1$ . In turn, any object which is non-negative and integrates to one can serve as a density. May  $\rho_t$  serve as a probability density?

## Theorem

*If  $b_t$  is bounded, then  $\rho_t[b]$  is a martingale and  $E\rho_t = 1$ .*

Proof. Let  $\tau_N := \inf(t \geq 0 : \rho_t \geq N)$ . Then clearly  $\int_0^t 1(s \leq \tau_N) b_s \rho_s dW_s$  is a martingale ("mart") because

$$E \int_0^t 1^2(s \leq \tau_N) b_s^2 \rho_s^2 ds \leq t \|b\|_B^2 N^2 < \infty.$$

So,  $E\rho_{t \wedge \tau_N} = 1 + E \int_0^{t \wedge \tau_N} b_s \rho_s dW_s = 1$ .

$$\rho_t[b] = \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right)$$

$$\rho_t = 1 + \int_0^t b_s \rho_s dW_s; \quad E \rho_{t \wedge \tau_N} = 1 + E \int_0^{t \wedge \tau_N} b_s \rho_s dW_s = 1$$

Moreover, by virtue of the martingale property for  $\rho_{t \wedge \tau_N}$ ,

$$E(\rho_{t \wedge \tau_N} | \mathcal{F}_s) = \rho_{s \wedge \tau_N}, \quad s < t.$$

Here due to the continuity of  $\rho_t$ ,

$$\tau_N \rightarrow \infty, \quad N \rightarrow \infty.$$

Therefore, the right hand side here tends to  $\rho_s$  as  $N \rightarrow \infty$ . What happens with the left hand side? We would show the martingale property of  $\rho_t$  if we knew that  $\rho_{t \wedge \tau_N}$  is uniformly integrable. Indeed, uniform integrability allows to use Lebesgue's analogue of the dominated convergence theorem *for conditional expectations*, under the U.I. condition instead of the domination assumption. *[This is a material for your homework: to repeat all limit theorems for conditional expectatoins.]*

$$\rho_t[b] = \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right)$$

$$\rho_t = 1 + \int_0^t b_s \rho_s dW_s; \quad E\rho_{t \wedge \tau_N} = 1 + E \int_0^{t \wedge \tau_N} b_s \rho_s dW_s = 1$$

So, it suffices to show that

$$E\rho_{t \wedge \tau_N}^2 \leq C$$

with some  $C$ , independent of  $N$ . We estimate,

$$\begin{aligned} E\rho_{t \wedge \tau_N}^2[b] &= E \exp\left(2 \int_0^{t \wedge \tau_N} b_s dW_s - \int_0^{t \wedge \tau_N} b_s^2 ds\right) \\ &= E \exp\left(\int_0^{t \wedge \tau_N} 2b_s dW_s - \frac{1}{4} \int_0^{t \wedge \tau_N} (2b)_s^2 ds\right) \\ &= E \exp\left(\int_0^{t \wedge \tau_N} 2b_s dW_s - \frac{1}{2} \int_0^{t \wedge \tau_N} (2b)_s^2 ds + \frac{1}{4} \int_0^{t \wedge \tau_N} (2b)_s^2 ds\right) \\ &\leq \exp\left(\frac{1}{4} t \|(2b)^2\|_B\right) E\rho_{t \wedge \tau_N}[2b] = \exp\left(\frac{1}{4} t \|(2b)^2\|_B\right) < \infty. \end{aligned}$$

Note that the right hand side here does not depend on  $N$ .

$$\rho_t = \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right)$$
$$\rho_t = 1 + \int_0^t b_s \rho_s dW_s$$

Thus, for any bounded  $b$  the stochastic exponential  $\rho_t$  is a (non-negative) martingale with  $E\rho_t = 1$ . Hence, this exponential may serve as a probability density. Let us define a new measure on  $\mathcal{F}$ ,

$$\tilde{P}(A) = P^{\rho_t}(A) := E\rho_t \mathbf{1}(A).$$

*[Homework: check that  $\tilde{P}$  is, indeed, a probability measure.]*

Can the boundedness of  $b$  for the martingale property of  $\rho$  be relaxed and how far? The most well-known is Novikov's condition

$$E \exp\left(\frac{1}{2} \int_0^t b_s^2 ds\right) < \infty.$$

There were preceding conditions by Gikhman and Skorokhod, and there are extensions due to Krylov. We will learn one small step towards these weaker conditions.

# Martingale property of $\rho_t = \exp(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds)$

Assumption  $E \exp(C \int_0^t b_s^2 ds) < \infty$

Let us return to the calculus establishing the uniform integrability of  $\rho_{\cdot \wedge T_N}$ ; we will try to improve it a bit. We have,

$$\begin{aligned} E \rho_{t \wedge T_N}^2 [b] &= E \exp\left(2 \int_0^{t \wedge T_N} b_s dW_s - \int_0^{t \wedge T_N} b_s^2 ds\right) \\ &= E \exp\left(\int_0^{t \wedge T_N} 2b_s dW_s - (4 - 3) \int_0^{t \wedge T_N} b_s^2 ds\right) \\ &\stackrel{CBS}{\leq} \left( E \exp\left(\int_0^{t \wedge T_N} 4b_s dW_s - 8 \int_0^{t \wedge T_N} b_s^2 ds\right) \right)^{\frac{1}{2}} \\ &\times \left( E \exp\left(6 \int_0^{t \wedge T_N} b_s^2 ds\right) \right)^{\frac{1}{2}} \leq \left( E \exp\left(6 \int_0^t b_s^2 ds\right) \right)^{\frac{1}{2}}. \end{aligned}$$

A conclusion: the condition  $E \exp(6 \int_0^t b_s^2 ds) < \infty$  suffices.



# Supermart property of $\rho_t = \exp(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds)$

Assumption  $P(\int_0^t b_s^2 ds < \infty) = 1$  (so that a SI  $\int_0^t b_s dW_s$  is defined)

Theorem (recall that  $\rho_t = 1 + \int_0^t b_s \rho_s dW_s$ )

*Under the assumption  $P(\int_0^t b_s^2 ds < \infty) = 1$  the process  $\rho$  is a supermartingale:  $\rho_{t_1} \geq E(\rho_{t_2} | \mathcal{F}_{t_1})$ ,  $\forall t_1 < t_2$ , &  $E\rho_t \leq 1$ .*

Proof. Return to the beginning of the proof of the last theorem. With a stopping time  $\tau_N := \inf(t \geq 0 : \rho_t \geq N)$ , the process  $\int_0^t \mathbf{1}(s \leq \tau_N) b_s \rho_s dW_s$  is a martingale, so,

$$1 + E\left(\int_0^{t_2 \wedge \tau_N} b_s \rho_s dW_s \middle| \mathcal{F}_{t_1}\right) = 1 + \int_0^{t_1 \wedge \tau_N} b_s \rho_s dW_s.$$

In other words,  $E(\rho_{t_2 \wedge \tau_N} | \mathcal{F}_{t_1}) = \rho_{t_1 \wedge \tau_N}$ . The supermart inequality follows from the Fatou lemma for conditional expectations  $E(\liminf_{N \rightarrow \infty} \rho_{t_2 \wedge \tau_N} | \mathcal{F}_{t_1}) \leq \liminf_{N \rightarrow \infty} \rho_{t_1 \wedge \tau_N}$ , since  $\rho_{s \wedge \tau_N} \rightarrow \rho_s$  due to continuity of  $\rho$ .

# Corollary

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## Corollary

*For any bounded adapted process  $b_t$ ,*

$$E \exp\left(\int_0^t b_s dW_s\right) < \infty.$$

# New measure $\tilde{P} \mapsto$ new WP $\tilde{W}$

Assume  $\rho$  is a mart on  $[0, t]$ ; define  $\tilde{W}_s := W_s - \int_0^s b_u du, s \leq t$ .

The next question is natural: we changed our measure; it is likely that  $W_s$  is no more a WP under this new measure; but is there a new WP instead? Igor Vladimirovich Girsanov proposed a new WP to be:

$$\tilde{W}_s := W_s - \int_0^s b_u du, \quad 0 \leq s \leq t.$$

## Theorem (Girsanov)

*Let  $b_t$  be bounded. Then  $\tilde{W}_s$  is a Wiener process on  $[0, t]$  under the measure  $\tilde{P} : d\tilde{P}/dP = \rho_t$ .*

As an immediate consequence, for any bounded Borel drift  $b(\cdot)$  we can construct a *weak* solution of an SDE

$$dX_t = b(X_t)dt + dW_t, \quad X_0 = x. \quad (3)$$

# Weak solution from Girsanov's theorem

By changing measure!

Denote  $X_s = W_s + x$  and

$$\tilde{W}_s = W_s - \int_0^s b(x + W_u) du, \quad s \leq t.$$

This is a new WP under the new probability measure

$$\frac{d\tilde{P}}{dP} = \rho_t := \exp\left(\int_0^t b(x + W_s) dW_s - \frac{1}{2} \int_0^t b^2(x + W_s) ds\right).$$

Then we have,

$$W_s = \tilde{W}_s + \int_0^s b(x + W_u) du, \quad s \leq t.$$

and therefore,  $X$  is a solution to the SDE with a new WP on  $[0, t]$ ,

$$X_s = x + \tilde{W}_s + \int_0^s b(X_u) du, \quad s \leq t.$$

# Proof of Girsanov's theorem about a new WP

$\tilde{W}_s := W_s - \int_0^s b_u du$ , and  $d\tilde{P} = \rho_t dP$  with  $\rho_t = \exp(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds)$

For the proof we need one lemma and the definition of a WP via its characteristic function, namely, for any

$0 = t_0 < t_1 < t_2 \dots < t_N$  and real values  $\lambda_j$ ,  $1 \leq j \leq N$ ,

$$\tilde{E} \exp\left(\sum_{j=0}^{N-1} i\lambda_j(\tilde{W}_{t_{j+1}} - \tilde{W}_{t_j})\right) = \exp\left(-\frac{1}{2} \sum_j \lambda_j^2 (t_{j+1} - t_j)\right).$$

## Lemma

Let  $\beta_t = \beta_t^1 + i\beta_t^2$  be a bounded adapted random process, where  $i = \sqrt{-1}$ . Then the (complex-valued) process

$$\rho_t[\beta] := \exp\left(\int_0^t \beta_s dW_s - \frac{1}{2} \int_0^t \beta_s^2 ds\right)$$

is a (complex-valued) martingale.

# Proof of Theorem

$$\rho_t[b] = \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right); \tilde{W}_s := W_s - \int_0^s b_u du$$

Let  $\lambda_s := i\lambda_j$  on  $[t_j, t_{j+1})$ , and  $B_s = b_s + \lambda_s$ . Then

$$\begin{aligned} \tilde{E} \exp\left(\sum_{j=0}^{N-1} i\lambda_j(\tilde{W}_{t_{j+1}} - \tilde{W}_{t_j})\right) &= E \rho_t[b] \exp\left(\sum_{j=0}^{N-1} i\lambda_j(\tilde{W}_{t_{j+1}} - \tilde{W}_{t_j})\right) \\ &= E \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right) \\ &\quad \times \exp\left(\sum_{j=0}^N i\lambda_j(W_{t_{j+1}} - W_{t_j} - \int_{t_j}^{t_{j+1}} b_u du)\right) \\ &= E \exp\left(\int_0^t B_s dW_s - \frac{1}{2} \int_0^t B_s^2 ds + \frac{1}{2} \int_0^t \lambda_s^2 ds\right). \end{aligned}$$

Indeed,

$$B_s^2 = (b_s + \lambda_s)^2 = b_s^2 + \lambda_s^2 + 2b_s\lambda_s.$$

# End of Proof of Girsanov's theorem

$$\lambda_s := i\lambda_j \text{ on } [t_{j+1} - t_j]; B_s = b_s + \lambda_s$$

But due to the last Lemma

$$E \exp\left(\int_0^t B_s dW_s - \frac{1}{2} \int_0^t B_s^2 ds\right) = 1$$

while  $\frac{1}{2} \int_0^t \lambda_s^2 ds$  is non-random and equals

$$\frac{1}{2} \int_0^t \lambda_s^2 ds = -\frac{1}{2} \sum_j \lambda_j^2 (t_{j+1} - t_j).$$

Therefore,

$$\tilde{E} \exp\left(\sum_{j=0}^{N-1} i\lambda_j (\tilde{W}_{t_{j+1}} - \tilde{W}_{t_j})\right) = \exp\left(-\frac{1}{2} \sum_j \lambda_j^2 (t_{j+1} - t_j)\right),$$

as required. Girsanov's theorem about a new WP under the (Girsanov's) change of measure is proved.

# Proof of Lemma

(Recall:)

## Lemma

Let  $\beta_t = \beta_t^1 + i\beta_t^2$  be a bounded adapted random process, where  $i = \sqrt{-1}$ . Then the (complex-valued) process

$$\rho_t[\beta] := \exp\left(\int_0^t \beta_s dW_s - \frac{1}{2} \int_0^t \beta_s^2 ds\right)$$

is a (complex-valued) martingale.

Proof. It suffices to check for any  $A \in \mathcal{F}_{t_1}$ ,  $t_2 > t_1$ , and a complex value  $z$ ,

$$\begin{aligned} & E1(A) \exp\left(\int_0^{t_2} (\beta_s^1 + z\beta_s^2) dW_s - \frac{1}{2} \int_0^{t_2} (\beta_s^1 + z\beta_s^2)^2 ds\right) \\ &= E1(A) \exp\left(\int_0^{t_1} (\beta_s^1 + z\beta_s^2) dW_s - \frac{1}{2} \int_0^{t_1} (\beta_s^1 + z\beta_s^2)^2 ds\right). \end{aligned}$$



We already know the equality for all real-valued  $z$ ,

$$\begin{aligned} & E1(A) \exp\left(\int_0^{t_2} (\beta_s^1 + z\beta_s^2) dW_s - \frac{1}{2} \int_0^{t_2} (\beta_s^1 + z\beta_s^2)^2 ds\right) \\ &= E1(A) \exp\left(\int_0^{t_1} (\beta_s^1 + z\beta_s^2) dW_s - \frac{1}{2} \int_0^{t_1} (\beta_s^1 + z\beta_s^2)^2 ds\right). \end{aligned}$$

Hence, the claim will be proved if we show that both sides are analytic functions of  $z$ . For the latter, it suffices to show that both sides are continuous in  $z$  and that their integrals along any closed bounded contours are equal to zero (Morera's theorem). Because of the analyticity of the expressions **under** the expectations and by Fubini's theorem (i.e., we can change the order of expectation and integration over the contour), we only need to show that for any  $R > 0$  and  $|z| \leq R$ , these expressions are bounded by an integrable r.v. independently of  $z$ .

Clearly, to show such domination we only need to care about the stochastic integrals (since Lebesgue's ones are bounded for  $|z| \leq R$ ). By virtue of the clever inequality for any  $\alpha, \beta \in \mathbb{R}$  with  $|\alpha| \leq |\beta|$ ,

$$\exp(\alpha) \leq \exp(\alpha) + \exp(-\alpha) \leq \exp(\beta) + \exp(-\beta),$$

we have,

$$\begin{aligned} \left| \exp\left(\int_0^{t_2} (\beta_s^1 + z\beta_s^2) dW_s\right) \right| &= \exp\left(\int_0^{t_2} (\beta_s^1 + \operatorname{Re}(z)\beta_s^2) dW_s\right) \\ &\leq \exp\left(\int_0^{t_2} (\beta_s^1 + R\beta_s^2) dW_s\right) + \exp\left(\int_0^{t_2} (\beta_s^1 - R\beta_s^2) dW_s\right). \end{aligned}$$

The latter expression is integrable independently of  $z$  (of course, for  $|z| \leq R$ ).

# Exponential inequality via stochastic exponential

$$\rho_t[b] = \exp\left(\int_0^t b_s dW_s - \frac{1}{2} \int_0^t b_s^2 ds\right)$$

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## Theorem (SI exp bounds)

*Let the adapted process  $b$ . be bounded. Then there exist  $C_1, C_2$  such that for any  $a > 0$  and for any  $T > 0$ ,*

$$P\left(\sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right| \geq a\right) \leq C_1 \exp(-a^2 / (C_2 T)).$$

The setting is  $d$ -dimensional;  $b$  here is a vector. It is also true for  $b$  matrices with some slight changes in constants.

## Corollary

*Under the same assumptions, for any  $T > 0 \exists \alpha > 0$  such that*

$$E \exp\left(\alpha \sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right|^2\right) < \infty.$$



# Proof

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For any  $\lambda$ ,  $\rho_t[\lambda b]$  is a continuous martingale. So, with any  $\lambda > 0$  by Bienaymé–Chebyshev–Markov’s inequality we have,

$$\begin{aligned} & P\left(\sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right| \geq a\right) \\ & \leq P\left(\sup_{0 \leq t \leq T} \int_0^t \lambda b_s dW_s \geq \lambda a\right) + P\left(\sup_{0 \leq t \leq T} \int_0^t (-\lambda) b_s dW_s \geq \lambda a\right) \\ & \leq e^{-\lambda a} E e^{\sup_{0 \leq t \leq T} \int_0^t \lambda b_s dW_s} + e^{-\lambda a} E e^{\sup_{0 \leq t \leq T} \int_0^t (-\lambda) b_s dW_s}. \end{aligned}$$

Consider each term separately and in the same manner.

# Proof, ctd.

In the middle of the calculus we use Doob's inequality:

$$\begin{aligned} e^{-\lambda a} E e^{\sup_{0 \leq t \leq T} \int_0^t \lambda b_s dW_s} &= e^{-\lambda a} E \sup_{0 \leq t \leq T} e^{\int_0^t \lambda b_s dW_s} \\ &= e^{-\lambda a} E \sup_{0 \leq t \leq T} \rho_t[\lambda b] e^{+\frac{1}{2} \int_0^t (\lambda b_s)^2 ds} \leq e^{-\lambda a + Ct\lambda^2} E \sup_{0 \leq t \leq T} \rho_t[\lambda b] \\ &\leq e^{-\lambda a + Ct\lambda^2} \sqrt{4E\rho_T^2[\lambda b]} \\ &= 2e^{-\lambda a + Ct\lambda^2} \left( E\rho_T[2\lambda b] \exp\left(\int_0^t (\lambda b_s)^2 ds\right) \right)^{1/2} \\ &\stackrel{\text{new } C}{\leq} 2e^{-\lambda a + Ct\lambda^2} (E\rho_T[2\lambda b])^{1/2} = 2 \exp(-\lambda a + Ct\lambda^2). \end{aligned}$$

Taking  $\inf_{\lambda > 0}$ , obtain with  $\lambda = a/(2Ct)$  the bound

$$e^{-\lambda a} E e^{\sup_{0 \leq t \leq T} \int_0^t \lambda b_s dW_s} \leq 2 \exp(-a^2/(4Ct)).$$

# Proof, ctd.

Check yourself that the other term

$$e^{-\lambda a} e^{\sup_{0 \leq t \leq T} \int_0^t (-\lambda) b_s dW_s}$$

admits the same bound,

$$e^{-\lambda a} E e^{\sup_{0 \leq t \leq T} \int_0^t (-\lambda b_s) dW_s} \leq 2 \exp(-a^2 / (4Ct)).$$

Overall, we obtain, as required,

$$\begin{aligned} P\left(\sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right| \geq a\right) &\leq e^{-\lambda a} E e^{\sup_{0 \leq t \leq T} \int_0^t \lambda b_s dW_s} \\ &\quad + e^{-\lambda a} e^{\sup_{0 \leq t \leq T} \int_0^t (-\lambda) b_s dW_s} \leq 4 \exp(-a^2 / (4Ct)). \end{aligned}$$

# Proof of Corollary

The idea is to use the bound with  $a^2 = z \geq 0$

$$\begin{aligned} P\left(\sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right|^2 \geq a^2\right) &= P\left(\sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right| \geq a\right) \\ &\leq 4 \exp(-a^2 / (4Ct)). \end{aligned}$$

Now integrate (in the middle by parts) with  $\alpha < (4Ct)^{-1}$ :

$$\begin{aligned} &E \exp\left(\alpha \sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right|^2\right) \\ &= \int_0^\infty \exp(\alpha z) d\left(-P\left(\sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right|^2 \geq z\right)\right) \\ &= 1 + \int_0^\infty P\left(\sup_{0 \leq t \leq T} \left| \int_0^t b_s dW_s \right|^2 \geq z\right) d \exp(\alpha z) \\ &\leq 1 + \alpha \int_0^\infty 4 \exp(-z[(4Ct)^{-1} - \alpha]) dz < \infty. \end{aligned}$$