

Girsanov Change of Measure: SDE Inference

Math653: SDEs and applications

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$$dX_t = b(X_t)dt + \sigma(X_t)dB_t; \quad X_0 = x_0 \in \mathbb{R}.$$

Motivating Questions

- How to estimate the drift b from path data $\{X_{[0,T]}^m\}$?
- How to classify if a path is from SDE with b_1 or SDE with b_2 ? (Hypothesis test)

Motivation: Likelihood on Path Space

Likelihood on Path Space

Warm-up: i.i.d. case

To estimate θ_0 from data $X^{(1)}, \dots, X^{(M)} \stackrel{\text{i.i.d.}}{\sim} X$ with pdf u_{θ_0} , we maximize the likelihood

$$\ell_M(\theta) = \frac{1}{M} \sum_{m=1}^M \log u_{\theta}(X^{(m)}).$$

Note: u_{θ} is defined relative to a Lebesgue measure.

Inference for SDEs

How to estimate b from path data $\{X_{[0,T]}^m\}$ of SDE?

$$dX_t = b(X_t) dt + \sigma(X_t) dB_t, \quad t \in [0, T],$$

What is the analogue of the likelihood $\ell_M(\theta)$ for SDEs?

What should the *density of the path* $X_{[0,T]}$ mean?

Pseudo-Likelihood from Euler–Maruyama

On a grid $t_k = k\Delta$,

$$X_{t_{k+1}} | X_{t_k} \approx \mathcal{N}(X_{t_k} + b(X_{t_k})\Delta, \sigma^2(X_{t_k})\Delta).$$

This suggests the approximate log-likelihood of $X_{t_0:t_L}$:

$$\log u(x_{0:L}) \approx -\frac{1}{2} \sum_k \left(\log \sigma^2(x_{t_k}) + \frac{(\Delta x_k - b(x_{t_k})\Delta)^2}{\sigma^2(x_{t_k})\Delta} \right).$$

Temptation: Let $\Delta \rightarrow 0$, $\Delta x_k / \Delta \rightarrow \dot{X}_t$. ↓

$$\mathcal{E}(b; X_{[0,T]}) = \int_0^T \frac{|\dot{X}_t - b(X_t)|^2}{2\sigma^2(X_t)} dt.$$

But \dot{X}_t does not exist a.s. (nowhere differentiable).

What is wrong with this argument?

Reference Law: Choose a reference path law, typically the zero-drift:

$$X_{t_{k+1}} | X_{t_k} \approx \mathcal{N}(X_{t_k}, \sigma^2(X_{t_k})\Delta).$$

A reference pdf of $X_{t_0:t_L}$: $u_0(x_{0:L})$

$$\log u_0(x_{0:L}) \approx -\frac{1}{2} \sum_k \left(\log \sigma^2(x_{t_k}) + \frac{(\Delta x_k)^2}{\sigma^2(x_{t_k})\Delta} \right).$$

The likelihood ratio of the drifted law versus the reference law:

$$\begin{aligned} \log \frac{u(x_{0:L})}{u_0(x_{0:L})} &\approx -\frac{1}{2} \sum_k \left(\frac{(\Delta x_k - b(x_{t_k})\Delta)^2}{\sigma^2(x_{t_k})\Delta} - \frac{(\Delta x_k)^2}{\sigma^2(x_{t_k})\Delta} \right) \\ &\approx -\frac{1}{2} \sum_k \left(\frac{b^2(x_{t_k})\Delta^2 - 2b(x_{t_k})\Delta x_k \Delta}{\sigma^2(x_{t_k})\Delta} \right) \\ &\xrightarrow{\Delta \rightarrow 0} - \int_0^T \frac{b^2(X_t)dt - 2b(X_t)dX_t}{2\sigma^2(X_t)} =: \mathcal{E}(b) \end{aligned}$$

Q: Ratio has limit. What is the limit of $u_0(x_{0:L})$ as $\Delta \rightarrow 0$?

Reference Law: The Right Likelihood Question

We cannot write the path pdf as the finite-dimensional case.

Denote it by \mathbb{P}^b for SDE with drift b .

Choose a reference path law \mathbb{P}^0 , and define the "ratio" between the two laws.

$$\frac{d\mathbb{P}^b}{d\mathbb{P}^0} \Big|_{\mathcal{F}_T},$$

where \mathcal{F}_T is the sigma-algebra generated by $X_{[0,T]}$.

Main message

Girsanov's theorem gives this Radon–Nikodym derivative when the diffusion is unchanged and only the drift is modified.

Change of Measure Basics

Change of Measure: What Replaces a Path Density?

Absolute continuity

$$\mathbb{Q} \ll \mathbb{P} \text{ on } \mathcal{F}_T \iff \mathbb{P}(A) = 0 \implies \mathbb{Q}(A) = 0 \quad \text{for all } A \in \mathcal{F}_T.$$

Radon–Nikodym derivative

Then there exists $Z_T \geq 0$, \mathcal{F}_T -measurable, such that

$$\mathbb{Q}(A) = \mathbb{E}_{\mathbb{P}}[\mathbf{1}_A Z_T], \quad Z_T = \left. \frac{d\mathbb{Q}}{d\mathbb{P}} \right|_{\mathcal{F}_T}.$$

- Interpretation: Z_T is the weight that converts \mathbb{P} -expectations into \mathbb{Q} -expectations.
- **Density process:** (Z_t) is a nonnegative \mathbb{P} -martingale:

$$Z_t := \mathbb{E}_{\mathbb{P}}[Z_T \mid \mathcal{F}_t], \quad Z_t = \left. \frac{d\mathbb{Q}}{d\mathbb{P}} \right|_{\mathcal{F}_t}.$$

- On path space, measure change is dynamic: the likelihood ratio itself evolves in time.

For SDE, what is the density process Z_t ?

Return to the change of measure for r.v., using the Radon–Nikodym derivative.

Z : a nonnegative random variable on $(\Omega, \mathcal{F}, \mathbb{P})$ with $\mathbb{E}_{\mathbb{P}}[Z] = 1$. Define a new measure \mathbb{Q} by

$$\mathbb{Q}(A) = \mathbb{E}_{\mathbb{P}}[\mathbb{1}_A Z], \quad A \in \mathcal{F}, \quad \text{i.e., } \frac{d\mathbb{Q}}{d\mathbb{P}} = Z.$$

Then for any random variable Y ,

$$\mathbb{E}_{\mathbb{Q}}[Y] = \mathbb{E}_{\mathbb{P}}[YZ].$$

Example: Let $X \sim \mathcal{N}(\lambda, 1)$ under \mathbb{P} . Define $Z = \exp(-\lambda X + \frac{1}{2}\lambda^2)$, then $\mathbb{E}_{\mathbb{P}}[Z] = 1$ and under \mathbb{Q} defined by $d\mathbb{Q}/d\mathbb{P} = Z$, the law of X is $\mathcal{N}(0, 1)$:

$$\begin{aligned} \mathbb{E}_{\mathbb{Q}}[e^{isX}] &= \mathbb{E}_{\mathbb{P}}[e^{isX} Z] = \mathbb{E}_{\mathbb{P}} \left[\exp \left((is - \lambda)X + \frac{1}{2}\lambda^2 \right) \right] \\ &= \exp \left(\lambda(is - \lambda) + \frac{1}{2}(is - \lambda)^2 + \frac{1}{2}\lambda^2 \right) = \exp \left(-\frac{1}{2}s^2 \right). \end{aligned}$$

Brownian motion: Let B_t be $N(0, t)$ under \mathbb{P} . Define $Z_t = \exp(-\lambda B_t - \frac{1}{2}\lambda^2 t)$, then $\mathbb{E}_{\mathbb{P}}[Z_t] = 1$, and $W_t = B_t + \lambda t$ is $N(0, t)$ under \mathbb{Q} defined by $d\mathbb{Q}/d\mathbb{P} = Z_t$.

Theorem (Girsanov's theorem for Brownian shift)

Let $(B_t)_{0 \leq t \leq T}$ be a Brownian motion under \mathbb{P} , adapted to the filtration $(\mathcal{F}_t)_{0 \leq t \leq T}$. Fix $\lambda \in \mathbb{R}$, and define

$$Z_t = \exp\left(-\lambda B_t - \frac{1}{2}\lambda^2 t\right), \quad 0 \leq t \leq T.$$

Then $(Z_t)_{0 \leq t \leq T}$ is a martingale under \mathbb{P} with $\mathbb{E}_{\mathbb{P}}[Z_t] = 1$, $0 \leq t \leq T$. Define a new probability measure \mathbb{Q} on \mathcal{F}_T by

$$\frac{d\mathbb{Q}}{d\mathbb{P}} \Big|_{\mathcal{F}_T} = Z_T.$$

Then, under \mathbb{Q} , the process $W_t := B_t + \lambda t$, $0 \leq t \leq T$, is a Brownian motion.

- Equivalently, under \mathbb{Q} , the process $(B_t)_{0 \leq t \leq T}$ has drift $-\lambda$, so that $B_t = W_t - \lambda t$.
- Z_t is a martingale by Itô's formula: $dZ_t = -\lambda Z_t dB_t$.