

## CSE 8803 CDS: Homework 5

Due April 19, '24 (11:59 pm ET) on Gradescope

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### 1 Problem 1 (10 points)

Here we would like to prove a foundational result in ergodic theory: Birkhoff ergodic theorem. Consider a map  $\varphi : M \rightarrow M$  with an ergodic measure,  $\mu$ . Prove that for any function  $f \in L^1(\mu)$ ,<sup>1</sup> its time average along an orbit,  $(1/T) \sum_{t \leq T} f(\varphi^t(x))$  converges to a constant, independent of  $x$  for  $\mu$  almost every  $x$ .

This result can be interpreted as a strong law of large numbers for the sequence  $f \circ \varphi^t(x)$ .

Hint: use Theorem 4.1.2 from KH. Then, any function that is constant along an orbit is constant almost everywhere for an ergodic system.

### 2 Problem 2 (20 points)

In this problem, we will implement an approximate score generative model (Song and Ermon 2019).

Part I Brownian motion or Wiener process is a popular choice for the forward process of an SGM. Simulate a Wiener process  $W_t, t \leq 1$ . Explain your code. (4 points)

Part II Let  $\rho_t$  be the solution of the Fokker Planck equation of  $dX_t = dW_t$ . Set  $\rho_0$  to be a bimodal Gaussian of your choice. What is  $\rho_t$ ? (3 points)

Part III The reverse SDE is defined as follows:

$$dY_t = \nabla \log \rho_{1-t}(Y_t) dt + dW_t, \quad (1)$$

with  $Y_0 \sim \rho_1$ . We will approximate the expression  $\nabla \log \rho_{1-t}(Y_t)$  with data obtained from the forward process. Simulate the reverse process until  $t = 1$ . Plot the distribution of  $Y_1$ . (5 points)

Part IV Do we expect  $Y_1$  to be distributed according to the target distribution (the bimodal Gaussian you set  $\rho_0$  to)? Why or why not? (8 points)

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<sup>1</sup> $L^1(\mu)$  is the space of functions with  $\int |f| d\mu \leq \infty$ .