Title: Supplementary material for computer practical class 2

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## The goal's (roughly speaking...)

Consider a Bayesian statistical model

$$\begin{cases} z_i | w & \stackrel{\text{ind}}{\sim} f(z_{1:n} | w), \ i = 1, ..., n \\ w & \sim f(w) \end{cases}$$
 (1)

Here: observables are  $\{z_i \in \mathcal{Z}\}_{i=1}^n$  and the unknown parameter is  $w \in \Theta$ .

The likelihood of the observables  $\{z_i \in \mathcal{Z}\}_{i=1}^n$  given the parameter  $w \in \Theta$  as

$$L_n(w) := f(z_{1:n}|w) = \prod_{i=1}^n f(z_i|w)$$

the posterior distribution density

$$f(w|z_{1:n}) = \frac{L_n(w) f(w)}{\int L_n(w) f(w) dw}$$

$$(2)$$

## Goal 1

Find mode of  $f(w|z_{1:n})$ , that is the Maximum aposteriori estimator (MAP) of w i.e.

$$w^* = \underset{w}{\operatorname{arg max}} \left\{ \log \left( f\left( w | z_{1:n} \right) \right) \right\}$$
$$= \underset{w \in \Theta}{\operatorname{arg max}} \left( \underbrace{\log \left( L_n\left( w \right) \right)}_{\text{(I)}} + \underbrace{\log \left( f\left( w \right) \right)}_{\text{(II)}} \right)$$

## Goal 2

Generate a random sample  $\{w^{(t)}\}$  from (2), in particular from

$$f_{\tau}\left(w|z_{1:n}\right) \propto \exp\left(\frac{1}{\tau} \prod_{i=1}^{n} f\left(z_{i}|w\right) f\left(w\right)\right)$$
 (3)

$$\propto \exp\left(\frac{1}{\tau}L_n\left(w\right)f\left(w\right)\right)$$
 (4)

## Gradient descent algorithm

Provide:

- A seed (initial value)  $w^{(0)}$ , m,  $\tau$
- Learning rate  $\eta_t : \mathbb{N} \to \mathbb{R}_+$ ; a positive non-increasing function such as  $\eta_t \to 0$ . E.g.
  - $-\eta_t = t_0/t$ , with  $t_0 > 0$  user specified
  - $-\eta_t = \text{something retatively tiny}$
- Temperature  $\tau > 0$

Stochastic Gradient Langevin Dynamics (SGLD) with learning rate  $\eta_t > 0$ , batch size m, and temperature  $\tau > 0$  is

- 1. Generate a random set  $\mathcal{J}^{(t)} \subseteq \{1,...,n\}^m$  of m indices from 1 to n with or without replacement, and set a  $\tilde{\mathcal{S}}_m = \{z_i; i \in \mathcal{J}^{(t)}\}.$
- 2. Compute

$$w^{(t+1)} = w^{(t)} + \eta_t \left( \frac{n}{m} \sum_{i \in J^{(t)}} \nabla \log f\left(z_i | w^{(t)}\right) + \nabla \log f\left(w^{(t)}\right) \right) + \sqrt{\eta_t} \sqrt{\tau} \epsilon_t \quad (5)$$

where  $\epsilon_{t} \stackrel{\text{iid}}{\sim} \mathcal{N}\left(0, I\right)$ .

3. Terminate if a termination criterion is satisfied; E.g.,  $t \leq T_{\text{max}}$  for a prespecified  $T_{\text{max}} > 0$ .