

Maximum Likelihood Estimate by Automatic Differentiation for Bernoulli Distribution

Weather
(w) $\begin{cases} \rightarrow \text{good} \\ \rightarrow \text{bad} \end{cases}$

$W \sim \text{Bernoulli}(\theta)$

have some data $D = \{G, G, G, B, G, B, \dots\}$
but no θ

log-likelihood
$$l(D; \theta) = \sum_{i=0}^{N-1} (w^{(i)} \log \theta + (1 - w^{(i)}) \log(1 - \theta))$$

$$\theta^* = \underset{\theta \in [0,1]}{\operatorname{argmax}} l(D; \theta)$$

$\frac{\partial l}{\partial \theta}$... tedious $\left(\begin{smallmatrix} \infty \\ \vdots \\ 0 \end{smallmatrix} \right)$

\rightarrow solve gradient-based optimization
by Automatic Differentiation

(we do not have to provide a
derivative)

$$\theta^* = \underset{\theta \in [0,1]}{\operatorname{argmin}} \underbrace{-l(D; \theta)}_{\substack{\text{negative log likelihood} \\ \hat{=} \text{"loss function"}}}$$